MINIMAL TIME FUNCTIONS AND THE SMALLEST INTERSECTING BALL PROBLEM GENERATED BY UNBOUNDED DYNAMICS

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Abstract: The smallest enclosing circle problem introduced in the 19th century by J. J. Sylvester \cite{Sylvester} asks for the circle of smallest radius enclosing a given set of finite points in the plane. An extension of the smallest enclosing circle problem called the smallest intersecting ball problem was considered in \cite{Yang1, Yang2}: given a finite number of nonempty closed subsets of a normed space, find a ball with the smallest radius that intersects all of the sets. In this paper we initiate the study of minimal time functions generated by unbounded dynamics and discuss their applications to extensions of the smallest intersecting ball problem. This approach continues our effort in applying convex and nonsmooth analysis to the well-established field of facility location.

Key words. Minimal time functions, subdifferential, subgradient method, the smallest intersecting ball problem.

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1 Introduction and Preliminaries

Given a finite number of points on the plane, the smallest enclosing circle problem asks for the smallest circle that encloses all of the points. This celebrated problem was introduced in the 19th century by the English mathematician J. J. Sylvester \cite{Sylvester} and is a simple example of continuous facility location problems. The reader is referred to \cite{Bose, Durocher, Smid} and the references therein for more recent results involving this problem in both theoretical and numerical aspects.

Let $X$ be a normed space and let $F \subseteq X$ be a nonempty closed convex subset of $X$. Given nonempty closed target sets $\Omega_i$ for $i = 1, \ldots, m$ and a nonempty closed constraint set $\Omega \subseteq X$, our goal in this paper is to study the following extended version of the smallest intersecting ball problem: find a point $\bar{x} \in \Omega$ and the smallest radius $r \geq 0$ (if they exist) such that the set $x + rF$ intersects all of the target sets. It is obvious that when $F$ is the Euclidean closed unit ball of the plane, the target sets under consideration are singletons, and the constraint set is the whole plane, this problem reduces to the smallest enclosing circle problem.

The so-called minimal time function below plays a crucial role in the study of the smallest intersecting ball problem. For a nonempty subset $Q \subseteq X$, define

$$T^F_Q(x) = \inf\{t \geq 0 : (x + tF) \cap Q \neq \emptyset\},$$

(1)
which signifies the minimal time for the point $x$ to reach the target set $Q$ following the dynamic $F$. General and generalized differentiation properties of the minimal time function (1) have been studied extensively in the literature; see, e.g. [5, 6, 8, 13, 15] and the references therein. When $F$ is the closed unit ball of $X$, the minimal time function becomes the familiar distance function

$$d(x; Q) = \inf \{||x - q|| : q \in Q\}.$$ 

In [4–6, 8], the minimal time function (1) in which $F$ is a closed bounded convex set that contains the origin as an interior point is considered. Further extensions to the case where the origin is not necessarily an interior point of $F$ have been considered in [10, 13]. However, to the best of our knowledge, the minimal time function (1) with $F$ being unbounded has not been addressed in the literature. In this paper, we are going to initiate the study of the minimal time function (1) for the case where $F$ is unbounded and consider applications to the corresponding extended version of the smallest intersecting ball problem.

Throughout the paper, we make the following standing assumption, unless otherwise specified:

$X$ is a normed space and $F \subseteq X$ is a nonempty closed convex set.

Under natural assumptions, the smallest intersecting ball problem generated by the dynamic $F$ with the target sets $\Omega_i$ for $i = 1, \ldots, m$ and the constraint set $\Omega$ can be modeled as the following optimization problem:

$$\text{minimize } T(x) \text{ subject to } x \in \Omega,$$

where

$$T(x) = \max\{T^F_{\Omega_i}(x) : i = 1, \ldots, m\}.$$ 

The functions involved in this problem are nondifferentiable with extended-real values in general, and they are convex when all of the sets therein are convex. Thus it is natural to use nonsmooth analysis and convex analysis in particular to study the problem.

Given a convex function $\varphi : X \to (-\infty, \infty]$ and a point $\bar{x} \in \text{dom } \varphi$, the subdifferential in the sense of convex analysis of the function $\varphi$ at $\bar{x}$ is defined by

$$\partial \varphi(\bar{x}) = \{x^* \in X^* : \langle x^*, x - \bar{x} \rangle \leq \varphi(x) - \varphi(\bar{x}) \text{ for all } x \in X\},$$

where the set $X^*$ is the dual of $X$ and the domain of the function $\varphi$ is

$$\text{dom } \varphi = \{x \in X : \varphi(x) < \infty\}.$$ 

An element $x^* \in \partial \varphi(\bar{x})$ is called a subgradient of the function $\varphi$ at $\bar{x}$. It is well known that subdifferentials of convex functions play an important role in convex optimization in both theoretical and numerical aspects. For instance, the generalized Fermat rule

$$\varphi \text{ has an absolute minimum at } \bar{x} \text{ if and only if } 0 \in \partial \varphi(\bar{x})$$

allows us to use subdifferentials as a tool to solve nonsmooth convex optimization problems. Under natural assumptions, the iterative sequence

$$x_{k+1} = x_k - \alpha_k x_k^*, \text{ with a chosen } x_1 \in X \text{ and } x_k^* \in \partial \varphi(x_k),$$

allows us to solve the problem.
converges to a minimizer of the function $\varphi$; see, e.g., [1] and the references therein.

Due to the involvement of the minimal time function (1) to the smallest intersecting ball problem, knowing subdifferential properties of this function provides important tools for the study of the problem in the convex setting. In fact, this has been shown by our recent papers [14, 16, 18] that generalized differentiation of the minimal time function (1) plays an important role in the study of set facility location problems including the smallest intersecting ball problem. It turns out that subdifferential properties of the minimal time function (1) have close connections to the asymptotic cone of $F$ and the normal cone to $Q$ at $\bar{x} \in Q$ defined by

$$N(\bar{x}; Q) = \{x^* \in X^* : \langle x^*, x - \bar{x} \rangle \leq 0 \text{ for all } x \in Q\}.$$ 

In this paper, we review important properties of asymptotic cones for convex sets in Section 2. Section 3 is devoted to general properties of the minimal time function (1), while generalized differentiation of this function in the convex case is addressed in Section 4. Finally, in Section 5, we provide applications of the results obtained to numerically solving the optimization problem (2) and its connection to the smallest intersecting ball problem. We review the literature and provide self-contained proofs for most of the results for the convenience of the reader.

Throughout the paper, $X^*$ denotes the dual space of $X$ and $\langle x^*, x \rangle = x^*(x)$ is the dual pair between an element $x^* \in X^*$ and an element $x \in X$.

## 2 Asymptotic Cones

In this section we review some facts about asymptotic cones that will be used in our results. For $x \in F$, the asymptotic cone of $F$ at $x$ is defined by

$$F_\infty(x) = \{d \in X : x + td \in F \text{ for all } t > 0\} = \{d \in X : x + td \in F \text{ for all } t \geq 0\}.$$ 

Another equivalent definition of $F_\infty(x)$ is given as

$$F_\infty(x) = \bigcap_{t>0} \frac{F - x}{t},$$

which shows that $F_\infty(x)$ is a cone and contains the origin, 0. In addition, $F_\infty(x)$ is closed and convex since $F$ is a closed convex set and the intersection of closed (convex) sets is closed (convex). Thus $F_\infty(x)$ is a closed convex cone.

**Proposition 2.1** For all $x_1, x_2 \in F$, we have

$$F_\infty(x_1) = F_\infty(x_2),$$

that is, the asymptotic cone does not depend on $x \in F$. Thus we will denote the asymptotic cone as $F_\infty$. 

3
Proof. For the proof of this statement in finite dimensions, the reader is referred [9]. The proof remains valid in infinite dimensions, and we include the details here for the convenience of the reader. Fix $x_1$ and $x_2$ in $F$. It suffices to show that

$$F_\infty(x_1) \subseteq F_\infty(x_2).$$

Let $d \in F_\infty(x_1)$. Fix any $t > 0$, we will show that $x_2 + td \in F$. Consider the sequence

$$x_n = x_1 + td + \left(1 - \frac{1}{n}\right)(x_2 - x_1),$$

which can be written as

$$x_n = \frac{1}{n}(x_1 + nd) + \left(1 - \frac{1}{n}\right)x_2.$$

We see that $x_n \in F$ for every $n$ because $d \in F_\infty(x_1)$ and $F$ is convex. In addition, $x_n \to x_2 + td$. Thus $x_2 + td \in F$ since $F$ is closed. Therefore, $d \in F_\infty(x_2)$. \qed

The following corollary is a direct consequence of Proposition 2.1.

**Corollary 2.1** Suppose $F$ contains the origin. Then

$$F_\infty = \bigcap_{t>0} tF.$$

**Proof.** Through the use of definitions of $F_\infty$ and Proposition 2.1, we have

$$F_\infty = F_\infty(0) = \bigcap_{t>0} \frac{F - 0}{t} = \bigcap_{t>0} \frac{1}{t}F = \bigcap_{t>0} tF$$

where $t = 1/\hat{t}$ for $\hat{t} > 0$. \qed

**Proposition 2.2** The following are equivalent:

(i) $d \in F_\infty$.

(ii) There exists a sequence $\{t_n\} \subseteq [0, \infty)$ such that $t_n \to 0$ and a sequence $\{f_n\} \subseteq F$ with $t_nf_n \to d$.

(iii) There exists a sequence $\{t_n\} \subseteq [0, \infty)$ such that $t_n \to 0$ and a sequence $\{f_n\} \subseteq F$ with $t_nf_n \xrightarrow{w} d$.

**Proof.** Let us prove “(i)⇒ (ii)”.

Suppose $d \in F_\infty$ and fix $\bar{x} \in F$. Then

$$\bar{x} + td \in F \text{ for all } t > 0.$$

In particular,

$$\bar{x} + nd \in F \text{ for all } n \in \mathbb{N}.$$

For each $n$, there exists $f_n \in F$ such that

$$\bar{x} + nd = f_n \text{ or equivalently } \frac{1}{n}\bar{x} + d = \frac{1}{n}f_n.$$

Let $t_n = \frac{1}{n}$. We easily see that $t_nf_n \to d$ as $n \to \infty$. \qed
The implication “(ii) ⇒ (iii)” is obvious since the strong convergence in $X$ implies the weak convergence. We finally prove “(iii) ⇒ (i)”. Assume that there exist sequences \( \{t_n\} \subseteq [0, \infty) \) and \( \{f_n\} \subseteq F \) such that \( t_n \to 0 \) and \( t_nf_n \xrightarrow{w} d \). Fix any \( x \in F \), and let \( t > 0 \). We will prove that \( d \in F_\infty \) by showing that

\[
x + td \in F \quad \text{for all } t > 0.
\]

One has \( 0 \leq t \cdot t_n < 1 \) when \( n \) is sufficiently large and \( t \cdot t_n \to 0 \). Thus

\[
(1 - t \cdot t_n)x + t \cdot t_nf_n \xrightarrow{w} x + td \quad \text{as } n \to \infty.
\]

The elements \((1 - t \cdot t_n)x + t \cdot t_nf_n\) are in \( F \) because \( F \) is convex, and since \( F \) is weakly closed, \( x + td \in F \). Therefore, \( d \in F_\infty \). \( \square \)

**Definition 2.1** The set \( F \) is called sequentially compact at \( 0 \) if for any sequence \( \{f_n\} \subseteq F \) such that \( f_n \xrightarrow{w} 0 \), one has \( f_n \to 0 \).

This property holds obviously in finite dimensions. Moreover, if \( X \) is a Hilbert space and \( F \) is a nonempty closed convex subset of the Fréchet normal cone to a set that is sequentially normally compact (see [12]) at the reference point, then \( F \) is automatically sequentially compact at \( 0 \). In infinite dimensions, it is not applicable to sets with nonempty interior.

**Proposition 2.3** If \( F \) is bounded, then \( F_\infty = \{0\} \). The converse holds if we suppose further that \( X \) is a reflexive Banach space and \( F \) is sequentially compact at the origin with \( 0 \in F \) or \( X \) is finite dimensional.

**Proof.** Suppose \( F \) is bounded and \( d \in F_\infty \). By Proposition 2.2, there exist sequences \( \{t_n\} \) such that \( t_n \to 0^+ \) and \( \{f_n\} \subseteq F \) such that

\[
t_nf_n \to d \quad \text{as } n \to \infty.
\]

Since \( F \) is bounded, the sequence \( \{t_nf_n\} \) converges to 0, so \( d = 0 \).

To prove the converse, we suppose by contradiction that \( F \) is unbounded and \( F_\infty = \{0\} \), while \( F \) is sequentially compact at the origin with \( 0 \in F \) or \( X \) is finite dimensional. Then there exits a sequence \( \{x_n\} \subseteq F \) with \( \|x_n\| \to \infty \). The sequence defined by

\[
d_n = \frac{x_n}{\|x_n\|}
\]

is bounded. So it has a subsequence (without relabeling) such that \( d_n \xrightarrow{w} d \in F \). Suppose first that \( F \) is sequentially compact at the origin with \( 0 \in F \). Since \( 0 \in F \) and \( F \) is convex, \( d_n \in F \) when \( n \) is sufficiently large. The sequentially compactness of \( F \) at 0 implies \( d \neq 0 \). Indeed, if \( d = 0 \), then \( d_n \to 0 \), but this is a contradiction since \( \|d_n\| = 1 \) for every \( n \). Fix any \( x \in F \), and let \( t > 0 \). For sufficiently large \( k \), one has

\[
u_k = \left(1 - \frac{t}{\|x_k\|}\right)x + \frac{t}{\|x_k\|}x_k \in F
\]

since \( F \) is convex. Since \( u_k \xrightarrow{w} x + td \) and \( F \) is weakly closed, \( x + td \in F \). Thus \( d \in F_\infty \), which is a contradiction. The proof is similar in the case \( X \) is finite dimensional. \( \square \)
3 General Properties of Minimal Time Functions with Unbounded Dynamics

In this section we study general properties of the minimal time function (1) that will be used in the next sections. These properties are also of independent interest.

It is obvious that if $x \in Q$, then $T^F_Q(x) = 0$. Moreover, $T^F_Q(x)$ is finite if there exists $t \geq 0$ such that

$$(x + tF) \cap Q \neq \emptyset.$$  

This holds in particular when $0 \in \text{int } F$. If no such $t$ exists, then $T^F_Q(x) = \infty$. Thus the minimal time function (1) is an extended real-valued function in general.

**Theorem 3.1** Suppose one of the following:

(i) $X$ is a reflexive Banach space, $Q$ is a bounded weakly closed set, and $F$ contains the origin 0.

(ii) $Q$ is a compact set and $F$ contains the origin 0.

Then

$T^F_Q(x) = 0$ if and only if $x \in Q - F_\infty$.

**Proof.** Let us prove the theorem under assumption (i). Suppose $T^F_Q(x) = 0$. Then for every $n \in \mathbb{N}$, there exists $t_n \geq 0$ such that

$$(x + t_n F) \cap Q \neq \emptyset,$$

and $\{t_n\}$ converges to 0. Then there exist $\omega_n \in Q$ and $f_n \in F$ for every $n \in \mathbb{N}$ with

$$x + t_n f_n = \omega_n.$$  

Since $Q$ is bounded and weakly closed in a reflexive Banach space, there exists a subsequence (without relabeling) such that the sequence $\{\omega_n\}$ converges weakly to $\omega \in Q$. Thus

$$t_n f_n \xrightarrow{w} \omega - x \text{ as } n \to \infty.$$  

It follows from Proposition 2.2 that $\omega - x \in F_\infty$, and hence $x \in Q - F_\infty$.

Let us prove the opposite implication. Fix any $x \in Q - F_\infty$. Then there exist $\omega \in Q$ and $d \in F_\infty$ such that

$$x = \omega - d.$$  

Since 0 $\in F$ and $d \in F_\infty$, then by Corollary 2.1, $n(\omega - x) = nd \in F$ for all $n \in \mathbb{N}$. Thus

$$\omega - x \in \frac{1}{n}F \text{ for all } n \in \mathbb{N}.$$  

This implies $(x + \frac{1}{n}F) \cap Q \neq \emptyset$ for all $n \in \mathbb{N}$. Thus $0 \leq T^F_Q(x) \leq \frac{1}{n}$ for all $n \in \mathbb{N}$. Therefore, $T^F_Q(x) = 0$. The proof of the theorem under assumption (ii) is similar. □

**Remark 3.1** The proof of Theorem 3.1 shows that, if $Q$ is a nonempty set (not necessarily bounded) and $0 \in F$, then $T^F_Q(x) = 0$ whenever $x \in Q - F_\infty$.  

\[ \text{Page 6} \]
The following example is a demonstration of Theorem 3.1.

Example 3.1 Let $F = \mathbb{R} \times [-1, 1] \subset \mathbb{R}^2$ and let $Q$ be the disk with center at $(1, 0)$ and radius $r = 1$. Then using the definition of $F_\infty$ in Corollary 2.1 we obtain

$$F_\infty = \mathbb{R} \times \{0\},$$

and $Q - F_\infty = \mathbb{R} \times [-1, 1]$. Thus $T_Q^F(x) = 0$ if and only if $x \in \mathbb{R} \times [-1, 1]$.

Let $\varphi : X \to (-\infty, \infty]$ be an extended real-valued function and let $\bar{x} \in \text{dom } \varphi$. Recall that $\varphi$ is lower semicontinuous at $\bar{x}$ if

$$\liminf_{x \to \bar{x}} \varphi(x) \geq \varphi(\bar{x}).$$

The function $\varphi$ is called lower semicontinuous if it is lower semicontinuous at every point of its domain.

Theorem 3.2 In the same setting of Theorem 3.1, the function $T_Q^F$ is lower semicontinuous.

Proof. Fix any $\bar{x} \in \text{dom } T_Q^F$ and fix a sequence $\{x_n\}$ that converges to $\bar{x}$. We will prove that

$$\liminf_{x \to \bar{x}} T_Q^F(x_n) \geq T_Q^F(\bar{x})$$

under the assumption (i) of Theorem 3.1. The inequality holds obviously if $\liminf T_Q^F(x_n) = \infty$. So we only need to consider the case where $\liminf T_Q^F(x_n) = \gamma \in [0, \infty)$. It suffices to show that $\gamma \geq T_Q^F(\bar{x})$. Without loss of generality, we can assume that $T_Q^F(x_n) \to \gamma$. By the definition of the minimal time function (1), there exists a sequence $\{t_n\} \subset [0, \infty)$ such that

$$T_Q^F(x_n) \leq t_n < T_Q^F(x_n) + \frac{1}{n} \quad \text{and} \quad (x_n + t_n F) \cap Q \neq \emptyset \quad \text{for all } n \in \mathbb{N}.$$ 

For every $n \in \mathbb{N}$, fix $f_n \in F$ and $\omega_n \in Q$ such that

$$x_n + t_n f_n = \omega_n,$$

and we can assume without loss of generality that $\omega_n \xrightarrow{w} \omega \in Q$. We will consider two cases: $\gamma > 0$ and $\gamma = 0$. If $\gamma > 0$, then

$$f_n \xrightarrow{w} \frac{\omega - \bar{x}}{\gamma} \in F.$$

Thus $(\bar{x} + \gamma F) \cap Q \neq \emptyset$, and hence $\gamma \geq T_Q^F(\bar{x})$.

Now we consider the case where $\gamma = 0$. In this case the sequence $\{t_n\}$ converges to 0, and

$$t_n f_n \xrightarrow{w} \omega - \bar{x}.$$

By Proposition 2.2, $\omega - \bar{x} \in F_\infty$. This implies $\bar{x} \in \Omega - F_\infty$. Employing Theorem 3.1, one has that $T_Q^F(\bar{x}) = 0$ and we also have $\gamma \geq T_Q^F(\bar{x})$. We have showed that $T_Q^F$ is lower semicontinuous under the assumption (i) of Theorem 3.1. The proof for the lower semicontinuity under the assumption (ii) is similar. □

7
Theorem 3.3 Suppose $Q \subseteq X$ is a nonempty convex set. Then the minimal time function (1) is convex.

Proof. Fix any $x_1, x_2 \in \text{dom } T^F_Q$ and let $\lambda \in (0, 1)$. We will show that
\begin{equation}
T^F_Q(\lambda x_1 + (1 - \lambda) x_2) \leq \lambda T^F_Q(x_1) + (1 - \lambda) T^F_Q(x_2).
\end{equation}
Let $\gamma_i = T^F_Q(x_i)$ for $i = 1, 2$. Given any $\epsilon > 0$, there exist $t_i$ for $i = 1, 2$ with
\begin{equation*}
\gamma_i \leq t_i < \gamma_i + \epsilon \text{ and } (x_i + t_i F) \cap Q \neq \emptyset.
\end{equation*}
Since both $F$ and $Q$ are convex, one has
\begin{equation*}
[\lambda x_1 + (1 - \lambda)x_2 + (\lambda t_1 + (1 - \lambda)t_2)F] \cap Q \neq \emptyset.
\end{equation*}
This implies
\begin{equation*}
T^F_Q(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda t_1 + (1 - \lambda)t_2 \leq \lambda T^F_Q(x_1) + (1 - \lambda)T^F_Q(x_2) + \epsilon.
\end{equation*}
Since $\epsilon$ is arbitrary, (1) holds and the proof is now complete. \hfill $\Box$

The Minkowski function associated with $F$ is defined by
\begin{equation}
\rho_F(x) = \inf\{t \geq 0 : x \in tF\}, \ x \in X.
\end{equation}
In contrast to the familiar definition of the Minkowski function, in this definition we only assume that $F$ is a nonempty closed convex subset of $X$.

Proposition 3.1 The Minkowski function $\rho_F$ is a positively homogenous and subadditive extended real-valued function. Suppose further that $0 \in F$. Then $\rho_F(x) = 0$ if and only if $x \in F_\infty$.

Proof. We first prove that $\rho_F$ is subadditive, that is,
\begin{equation}
\rho_F(x_1 + x_2) \leq \rho_F(x_1) + \rho_F(x_2) \text{ for all } x_1, x_2 \in X.
\end{equation}
The inequality holds obviously if the right side is infinity, that means $x_1 \notin \text{ dom } \rho_F$ or $x_2 \notin \text{ dom } \rho_F$. Suppose $x_1, x_2 \in \text{ dom } \rho_F$. Fix any $\epsilon > 0$. Then there exist $t_1, t_2 \geq 0$ such that
\begin{equation*}
\rho_F(x_1) \leq t_1 < \rho_F(x_1) + \epsilon, \ \rho_F(x_2) \leq t_2 < \rho_F(x_2) + \epsilon,
\end{equation*}
and
\begin{equation*}
x_1 \in t_1 F, \ x_2 \in t_2 F.
\end{equation*}
Since $F$ is convex,
\begin{equation*}
x_1 + x_2 \in t_1 F + t_2 F \subseteq (t_1 + t_2) F.
\end{equation*}
Thus $\rho_F(x_1 + x_2) \leq t_1 + t_2 < \rho_F(x_1) + \rho_F(x_2) + 2\epsilon$. This implies (3).
The fact that $\rho_F$ is positive homogeneous follows from the following analysis. For $\alpha > 0$, one has
\[
\rho_F(\alpha x) = \inf\{t \geq 0 : \alpha x \in t F\} \\
= \inf\{t \geq 0 : x \in \frac{t}{\alpha} F\} \\
= \alpha \inf\{\frac{t}{\alpha} \geq 0 : x \in \frac{t}{\alpha} F\} = \alpha \rho_F(x).
\]
Now, we will prove the second statement. Let $Q = \{0\}$. We have $\rho_F(x) = T_Q F(x)$. Since $0 \in F$, by Theorem 3.1, $\rho_F(x) = 0$ if and only if $x \in Q - (-F)_\infty = F_\infty$. The proof is now complete. □

Recall that a function $\varphi : X \to \mathbb{R}$ is said to be $\ell$–Lipschitz if
\[|\varphi(x) - \varphi(y)| \leq \ell ||x - y|| \text{ for all } x, y \in X.\]

**Proposition 3.2** Suppose $0 \in \text{int} F$. Define
\[\ell = \inf\left\{\frac{1}{r} : B(0; r) \subset F, r > 0\right\}.
\]
Then $\rho_F$ is an $\ell$–Lipschitz function. In particular, $\rho_F(x) \leq \ell ||x||$ for all $x \in X$.

**Proof.** It is clear that $\ell$ is a nonnegative real number since $0 \in \text{int} F$. Let $r > 0$ satisfy $B(0; r) \subset F$. Then
\[
\rho_F(x) = \inf\{t \geq 0 : x \in t F\} \leq \inf\{t \geq 0 : x \in t B(0; r)\} = \inf\{t \geq 0 : ||x|| \leq rt\} = \frac{||x||}{r}.
\]
This implies $\rho_F(x) \leq \ell ||x||$. Since $\rho_F$ is subadditive, for any $x, y \in X$, one has
\[
\rho_F(x) - \rho_F(y) \leq \rho_F(x - y) \leq \ell ||x - y||.
\]
This implies the $\ell$–Lipschitz property of $\rho_F$. □

Using the definitions of the minimal time function (1) and the Minkowski function (2), we can establish a relationship between these functions in the proposition below.

**Proposition 3.3** Let $Q \subseteq X$ be a nonempty closed set. Then
\[
T_Q F(x) = \inf\{\rho_F(\omega - x) : \omega \in Q\}.
\]  
(4)

**Proof.** Let us first consider the case where $T_Q F(x) = \infty$. Then the set
\[
\{t \geq 0 : (x + t F) \cap Q \neq \emptyset\} = \emptyset.
\]
It follows that for every $\omega \in Q$, the set
\[
\{t \geq 0 : \omega - x \in t F\} = \emptyset.
\]
Thus $\rho_F(\omega - x) = \infty$ for all $\omega \in Q$, and hence the right side of (4) is also infinity.
Now, suppose $T^F_Q(x) < \infty$. Fix any $t \geq 0$ such that

$$(x + tF) \cap Q \neq \emptyset.$$  

Then there exist $f \in F$ and $\omega \in Q$ with $x + tf = \omega$. Thus $\omega - x \in tF$, and hence $\rho_F(\omega - x) \leq t$. This implies

$$\inf_{\omega \in Q} \rho_F(\omega - x) \leq t.$$  

It follows by the definition of the minimal time function (1) that

$$\inf_{\omega \in Q} \rho_F(\omega - x) \leq T^F_Q(x) < \infty.$$  

Let $\gamma = \inf_{\omega \in Q} \rho_F(\omega - x)$. Then for any $\varepsilon > 0$, there exists $\omega \in Q$ with

$$\rho_F(\omega - x) < \gamma + \varepsilon.$$  

Employing the definition of the Minkowski function (2), one finds $t \in [0, \infty)$ with

$$\omega - x \in tF \text{ and } t < \gamma + \varepsilon.$$  

Thus $\omega \in x + tF$, and hence

$$T^F_Q(x) \leq t < \gamma + \varepsilon.$$  

Since $\varepsilon$ is arbitrary, one has

$$T^F_Q(x) \leq \gamma = \inf_{\omega \in Q} \rho_F(\omega - x).$$  

The proof is now complete.$\square$

Employing the $\ell$–Lipschitz property of the Minkowski function (2) from Proposition 3.2 and the representation of the minimal time function (1) in Proposition 3.3, we obtain the result below.

**Theorem 3.4** Let $\ell$ be defined in Proposition 3.2. Suppose $0 \in \text{int } F$. Then the minimal time function $T^F_Q$ is $\ell$–Lipschitz.

**Proof.** Let $x, y \in X$. Then using the subadditivity of $\rho_F$ from Proposition 3.1 and using Proposition 3.2, one has

$$\rho_F(\omega - x) \leq \rho_F(\omega - y) + \rho_F(y - x) \leq \rho_F(\omega - y) + \ell ||y - x|| \text{ for all } \omega \in Q.$$  

It follows from Proposition 3.3 that

$$T^F_Q(x) \leq T^F_Q(y) + \ell ||x - y||.$$  

This implies the $\ell$–Lipschitz property of $T^F_Q$. $\square$
4 Convex Subdifferentials of Minimal Time Functions with Unbounded Dynamics

In this section, we discuss properties of the convex subdifferentials of the minimal time function (1) where the target set $Q$ is convex. In this case the minimal time function (1) is also convex by Theorem 3.3. Consider the following extended real-valued function defined on $X^*$:

$$\sigma_F(x^*) = \sup\{\langle x^*, f \rangle : f \in F\}.$$ 

**Theorem 4.1** Consider the minimal time function (1) with a nonempty convex target set $Q$. For $\bar{x} \in Q$, one has

$$\partial T_F^Q(\bar{x}) = N(\bar{x}; Q) \cap C^*,$$

where

$$C^* = \{x^* \in X^* : \sigma_F(-x^*) \leq 1\}.$$ 

Suppose further that $0 \in F$. Then

$$\partial T_F^Q(\bar{x}) = N(\bar{x}; Q - F_{\infty}) \cap C^* = N(\bar{x}; Q) \cap C^*.$$ 

**Proof.** Fix any $x^* \in \partial T_F^Q(\bar{x})$. One has

$$\langle x^*, x - \bar{x} \rangle \leq T_F^Q(x) - T_F^Q(\bar{x}) \text{ for all } x \in X. \quad (5)$$

Since $T_F^Q(x) = 0$ for all $x \in Q$, one has

$$\langle x^*, x - \bar{x} \rangle \leq 0 \text{ for all } x \in Q.$$ 

This implies $x^* \in N(\bar{x}; Q)$. For any $f \in F$ and $t > 0$, it follows from (5) that

$$\langle x^*, (\bar{x} - tf) - \bar{x} \rangle \leq T_F^Q(\bar{x} - tf) \leq t.$$ 

Note that the last inequality holds since $((\bar{x} - tf) + tF) \cap Q \neq \emptyset$. Thus

$$\langle x^*, -f \rangle \leq 1 \text{ for all } f \in F,$$

so $x^* \in C^*$. Thus $\partial T_F^Q(\bar{x}) \subset N(\bar{x}; Q) \cap C^*$.

Now, we will show that

$$N(\bar{x}; Q) \cap C^* \subset \partial T_F^Q(\bar{x}).$$ 

Fix any $x^* \in N(\bar{x}; Q) \cap C^*$. Then $\langle x^*, -f \rangle \leq 1$ for all $f \in F$ and

$$\langle x^*, x - \bar{x} \rangle \leq 0 \text{ for all } x \in Q.$$ 

Fix any $u \in \text{ dom } T_F^Q$. For any $\varepsilon > 0$, there exist $t \in [0, \infty)$, $f \in F$, and $\omega \in Q$ with

$$T_F^Q(u) \leq t < T_F^Q(u) + \varepsilon \text{ and } u + tf = \omega.$$
Then
\[ \langle x^*, u - \bar{x} \rangle = \langle x^*, u - \omega \rangle + \langle x^*, \omega - \bar{x} \rangle \]
\[ \leq \langle x^*, -tf \rangle \leq t < T^F_Q(u) + \varepsilon = T^F_Q(u) - T^F_Q(\bar{x}) + \varepsilon. \]

This implies (5) since \( \varepsilon \) is arbitrary.

Let us finally prove that
\[ \partial T^F_Q(\bar{x}) = N(\bar{x}; Q - F_\infty) \cap C^* = N(\bar{x}; Q) \cap C^*, \]
under the additional assumption. Obviously, \( N(\bar{x}; Q - F_\infty) \cap C^* \subset N(\bar{x}; Q) \cap C^* \) since \( Q \subset Q - F_\infty \). Fix \( x^* \in N(\bar{x}; Q) \cap C^* \) and any \( x \in Q - F_\infty \). One has \( x = \omega - d \) for \( \omega \in Q \) and \( d \in F_\infty \). Since \( 0 \in F \), one has \( td \in F \) for all \( t > 0 \). This implies \( \langle x^*, -td \rangle \leq 1 \) for all \( t > 0 \). Thus \( \langle x^*, -d \rangle \leq 0 \). Then
\[ \langle x^*, x - \bar{x} \rangle = \langle x^*, \omega - d - \bar{x} \rangle = \langle x^*, \omega - \bar{x} \rangle + \langle x^*, -d \rangle \leq 0. \]

Therefore, \( x^* \in N(\bar{x}; Q - F_\infty) \cap C^* \). The proof is now complete. \( \square \)

Let us now consider the case where \( \bar{x} \in Q - F_\infty \). In fact, similar results like those in Theorem 4.1 hold true in this case.

**Theorem 4.2** Consider the minimal time function (1) with a nonempty convex target set \( Q \). Suppose \( \bar{x} \in Q - F_\infty \) and \( 0 \in F \). Then
\[ \partial T^F_Q(\bar{x}) = N(\bar{x}; Q - F_\infty) \cap C^*. \]

**Proof.** Fix any \( x^* \in \partial T^F_Q(\bar{x}) \). Using (5) and the fact that \( T^F_Q(x) = 0 \) for all \( x \in Q - F_\infty \) from Remark 3.1, one has that \( x^* \in N(\bar{x}; Q - F_\infty) \). Since \( \bar{x} \in Q - F_\infty \), one has \( \bar{x} = \bar{\omega} - d \), where \( \bar{\omega} \in Q \) and \( d \in F_\infty \). Fix any \( f \in F \). For any \( t > 0 \), let \( x_t = (\bar{x} + d) - tf = \bar{\omega} - tf \). Then \( (x_t + tf) \cap Q \neq \emptyset \). Thus
\[ \langle x^*, x_t - \bar{x} \rangle \leq T^F_Q(x_t) \leq t, \]
or equivalently,
\[ \langle x^*, d - tf \rangle \leq t. \]

Thus
\[ \langle x^*, \frac{d}{t} - f \rangle \leq 1 \] for all \( t > 0 \).

This implies \( \langle x^*, -f \rangle \leq 1 \) by letting \( t \to \infty \), and hence \( x^* \in C^* \). Thus we have shown that
\[ \partial T^F_Q(\bar{x}) \subset N(\bar{x}; Q - F_\infty) \cap C^*. \]

Now fix any \( x^* \in N(\bar{x}; Q - F_\infty) \cap C^* \). Fix any \( u \in \text{dom } T^F_Q \). Then there exist \( t \in [0, \infty) \), \( \omega \in Q \) and \( f \in F \) such that
\[ T^F_Q(u) \leq t < T^F_Q(u) + \varepsilon \text{ and } u + tf = \omega. \]

Since \( Q \subset Q - F_\infty \), one has
\[ \langle x^*, u - \bar{x} \rangle = \langle x^*, \omega - \bar{x} \rangle + t\langle x^*, -f \rangle \leq t < T^F_Q(u) + \varepsilon = T^F_Q(u) - T^F_Q(\bar{x}) + \varepsilon. \]
This again implies $x^* \in \partial \mathcal{T}_Q^F(\bar{x})$, and thus $N(\bar{x}; Q - F_\infty) \cap C^* \subset \partial \mathcal{T}_Q^F(\bar{x})$. The proof is now complete.

In what follows we are going to discuss the subdifferential in the sense of convex analysis for the minimal time function (1) when the reference point does not belong to $Q - F_\infty$. Our development is inspired by the early work on subdifferentials of distance functions; see [2, 3, 11]. For any $r > 0$, define

$$Q_r := \{x \in X : \mathcal{T}_Q^F(x) \leq r\}.$$

**Lemma 4.1** Let $Q$ be a nonempty set and let $r > 0$. Suppose $x \notin Q_r$ with $\mathcal{T}_Q^F(x) < \infty$. Then

$$\mathcal{T}_Q^F(x) = \mathcal{T}_{Q_r}^F(x) + r.$$

**Proof.** Let $r > 0$. Since $Q \subset Q_r$, one has $\mathcal{T}_{Q_r}^F(u) = \mathcal{T}_Q^F(u) \leq r$. Fix any $t \geq 0$ such that

$$(x + tF) \cap Q_r \neq \emptyset.$$  

Then there exist $f_1 \in F$ and $u \in Q_r$ such that $x + tf_1 = u$. Since $u \in Q_r$, then $\mathcal{T}_Q^F(u) \leq r$. Then for any $\varepsilon > 0$, there exists $s > 0$ such that $s < r + \varepsilon$. Consequently, there exist $\omega \in Q$ and $f_2 \in F$ such that $u + sf_2 = \omega$. Thus

$$\omega = u + sf_2 = (x + tf_1) + sf_2 \in x + (t + s)F$$

since $F$ is convex. This implies

$$\mathcal{T}_Q^F(x) \leq t + s \leq t + r + \varepsilon.$$  

Since $\varepsilon > 0$ is arbitrary, $\mathcal{T}_Q^F(x) \leq t + r$, and hence

$$\mathcal{T}_Q^F(x) \leq \mathcal{T}_{Q_r}^F(x) + r.$$  

Let us prove the opposite inequality. Let $\gamma = \mathcal{T}_Q^F(x)$. Then $r < \gamma$ since $x \notin Q_r$. For any $\varepsilon > 0$, there exist $t \in [0, \infty)$, $f \in F$, and $\omega \in Q$ with

$$\gamma \leq t < \gamma + \varepsilon \text{ and } x + tf = \omega.$$  

One has

$$\omega = x + tf = x + (t - r)f + r f \in x + (t - r)f + rF.$$  

Thus

$$\mathcal{T}_Q^F(x + (t - r)f) \leq r.$$  

So $x + (t - r)f \in Q_r$; in addition $x + (t - r)f \in x + (t - r)F$. Hence $\mathcal{T}_{Q_r}^F(x) \leq t - r \leq \gamma - r + \varepsilon$. Since $\varepsilon > 0$ is arbitrary, one has

$$r + \mathcal{T}_{Q_r}^F \leq \gamma = \mathcal{T}_Q^F(x).$$  

The proof is now complete.\qed
Lemma 4.2 Let \( Q \) be a nonempty set and let \( x \in \text{dom } T^F_Q \). For \( t \geq 0 \) and any \( f \in F \), one has

\[
T^F_Q(x - tf) \leq T^F_Q(x) + t.
\]

Proof. For any \( \varepsilon > 0 \), there exists \( s \geq 0 \) such that

\[
T^F_Q(x) \leq s < T^F_Q(x) + \varepsilon \quad \text{and} \quad (x + sF) \cap Q \neq \emptyset.
\]

Then \((x - tf + tF + sF) \cap Q \neq \emptyset\), and hence \((x - tf + (t + s)F) \cap Q \neq \emptyset\). Thus

\[
T^F_Q(x - tf) \leq t + s \leq T^F_Q(x) + t + \varepsilon.
\]

The conclusion follows by letting \( \varepsilon \to 0 \).

\[\Box\]

Theorem 4.3 Let \( X \) be reflexive Banach space and let \( 0 \in F \). Consider the minimal time function (1) with a nonempty closed bounded convex target set \( Q \). For \( \bar{x} \not\in Q - F_\infty \), one has

\[
\partial T^F_Q(\bar{x}) = N(\bar{x}; Q) \cap S^*,
\]

where

\[
S^* = \{ x^* \in X^* : \sigma_F(-x^*) = 1 \} \quad \text{and} \quad r = T^F_Q(\bar{x}) > 0.
\]

Proof. Fix any \( x^* \in \partial T^F_Q(\bar{x}) \). Following the proof of Theorem 4.1, see that \( \sigma_F(-x^*) \leq 1 \) and \( x^* \in N(\bar{x}; Q) \). Let us show that \( \sigma_F(-x^*) = 1 \). One has

\[
\langle x^*, x - \bar{x} \rangle \leq T^F_Q(x) - T^F_Q(\bar{x}) \quad \text{for all } x \in X.
\]

(6)

Fix any \( \varepsilon \in (0, r) \). There exist \( t \in \mathbb{R}, f \in F \) and \( \omega \in Q \) such that

\[
r \leq t < r + \varepsilon^2 \quad \text{and} \quad \omega = \bar{x} + tf.
\]

We can write \( \omega = \bar{x} + \varepsilon f + (t - \varepsilon)f \). So \( T^F_Q(\bar{x} + \varepsilon f) \leq t - \varepsilon \). Applying (6) for \( x = \bar{x} + \varepsilon f \), one has

\[
\langle x^*, \varepsilon f \rangle \leq T^F_Q(\bar{x} + \varepsilon f) - T^F_Q(\bar{x}) \leq t - \varepsilon - r \leq \varepsilon^2 - \varepsilon.
\]

This implies

\[
1 - \varepsilon \leq \langle -x^*, f \rangle \leq \sigma_F(-x^*).
\]

So \( \sigma_F(-x^*) \geq 1 \) by letting \( \varepsilon \to 0 \). Therefore, \( x^* \in S^* \), and thus \( \partial T^F_Q(\bar{x}) \subset N(\bar{x}; Q) \cap S^* \).

Fix any \( x^* \in N(\bar{x}; Q) \) such that \( \sigma_F(-x^*) = 1 \). We are going to show that (6) is satisfied. By Theorem 4.1, \( x^* \in \partial T^F_Q(\bar{x}) \). Thus

\[
\langle x^*, x - \bar{x} \rangle \leq T^F_Q(x) \quad \text{for all } x \in X.
\]

(6)

Fix any \( x \in X \). In the case \( t := T^F_Q(x) > r \), one has \( T^F_Q(x) - r = T^F_Q(x) \) by Lemma 4.1. Thus (6) holds. Suppose \( t \leq r \). For any \( \varepsilon > 0 \), choose \( f \in F \) such that \( \langle x^*, -f \rangle > 1 - \varepsilon \). By Lemma 4.2, \( T^F_Q(x - (r - t)f) \leq r \). So \( x - (r - t)f \in Q \). Since \( x^* \in N(\bar{x}; Q) \), one has

\[
\langle x^*, x - (r - t)f - \bar{x} \rangle \leq 0.
\]

This implies

\[
\langle x^*, x - \bar{x} \rangle \leq \langle x^*, f \rangle(r - t) \leq (1 - \varepsilon)(t - r) = (1 - \varepsilon)(T^F_Q(x) - T^F_Q(\bar{x})).
\]

Since \( \varepsilon \) is arbitrary, (6) holds. Therefore, \( x^* \in \partial T^F_Q(\bar{x}) \).

\[\Box\]
Theorem 4.4 Let $X$ be reflexive Banach space and let $0 \in F$. Consider the minimal time function (1) with a nonempty closed bounded convex target set $Q$. Suppose $\bar{x} \notin Q - F_\infty$ and $0 \in F$. Then

$$\partial T_Q^F(\bar{x}) = [-\partial \rho_F(\bar{\omega} - \bar{x})] \cap N(\bar{\omega}; Q)$$

for any $\bar{\omega} \in \Pi_F(\bar{x}; Q)$, where

$$\Pi_F(\bar{x}; Q) = \{ \omega \in Q : T_Q^F(\bar{x}) = \rho_F(\omega - \bar{x}) \}.$$

Proof. Fix any $x^* \in \partial T_Q^F(\bar{x})$ and $\bar{\omega} \in \Pi_F(\bar{x}; Q)$. Then $T_Q^F(\bar{x}) = \rho_F(\bar{\omega} - \bar{x})$ and

$$\langle x^*, x - \bar{x} \rangle \leq T_Q^F(x) - T_Q^F(\bar{x})$$

for all $x \in X$. (7)

For any $\omega \in Q$, one has

$$\langle x^*, \omega - \bar{\omega} \rangle = \langle x^*, (\omega - \bar{\omega} + \bar{x}) - \bar{x} \rangle$$

$$\leq T_Q^F(\omega - \bar{\omega} + \bar{x}) - T_Q^F(\bar{x})$$

$$\leq \rho_F(\omega - (\omega - \bar{\omega} + \bar{x})) - \rho_F(\bar{\omega} - \bar{x}) = 0.$$  

This implies $x^* \in N(\bar{\omega}; Q)$. Let $\bar{u} = \bar{\omega} - \bar{x}$. For any $t \in (0, 1)$ and $u \in X$, applying (7) with $x = \bar{x} - t(u - \bar{u})$, one has

$$\langle x^*, -t(u - \bar{u}) \rangle \leq T_Q^F(\bar{x} - t(u - \bar{u})) - T_Q^F(\bar{x})$$

$$\leq \rho_F(\bar{\omega} - (\bar{x} - t(u - \bar{u}))) - \rho_F(\bar{\omega} - \bar{x})$$

$$= \rho_F(\bar{u} + t(u - \bar{u})) - \rho_F(\bar{u})$$

$$= \rho_F(tu + (1 - t)\bar{u}) - \rho_F(\bar{u})$$

$$\leq t \rho_F(u) + (1 - t)\rho_F(\bar{u}) - \rho_F(\bar{u})$$

$$= t(\rho_F(u) - \rho_F(\bar{u})).$$

It follows that

$$\langle -x^*, u - \bar{u} \rangle \leq \rho_F(u) - \rho_F(\bar{u})$$

for all $u \in X$. Thus $-x^* \in \partial \rho_F(\bar{u})$, and hence $x^* \in -\partial \rho_F(\bar{\omega} - \bar{x})$. The inclusion “$\subset$” has been proved. Let us prove the opposite inclusion. We have that

$$\rho_F(x) = \inf \{ t \geq 0 : x \in tF \} = \inf \{ t \geq 0 : (-x + tF) \cap O \neq \emptyset \} = T_Q^F(-x),$$

where $O = \{ \emptyset \}$. For any $\bar{\omega} \in Q$ one has, $\bar{x} - \bar{\omega} \notin O - F_\infty = -F_\infty$ since $\bar{x} \notin Q - F_\infty$. From $-x^* \in \partial \rho_F(\bar{\omega} - \bar{x})$, one has $x^* \in S^*$ by Theorem 4.3. Thus we only need to show that

$$[-\partial \rho_F(\bar{\omega} - \bar{x})] \cap N(\bar{\omega}; Q) \subset N(\bar{x}; Q_r),$$

and the conclusion will follow from Theorem 4.3.

To proceed, pick any $x \in Q$, and for an arbitrary small $\varepsilon > 0$. Then there exist $t < r + \varepsilon$, $f \in F$, and $\omega \in Q$ with $\omega = x + tf$. Thus $\langle -x^*, f \rangle \leq \sigma_F(-x^*) = 1$ and

$$\langle x^*, x - \bar{x} \rangle = \langle x^*, \omega - tf - \bar{x} \rangle$$

$$= t(-x^*, f) + \langle x^*, \omega - \bar{\omega} \rangle + \langle x^*, \bar{\omega} - \bar{x} \rangle$$

$$\leq t + \langle x^*, \omega - \bar{\omega} \rangle + \langle x^*, \bar{\omega} - \bar{x} \rangle$$

$$\leq T_Q^F(\bar{x}) + \varepsilon + \langle x^*, \omega - \bar{\omega} \rangle + \langle x^*, \bar{\omega} - \bar{x} \rangle.$$
We have \( \langle x^*, \omega - \bar{\omega} \rangle \leq 0 \) due to \( x^* \in N(\bar{\omega}; Q) \) and
\[
\langle x^*, \omega - \bar{x} \rangle = \langle -x^*, 0 - (\omega - \bar{x}) \rangle \leq \rho_F(0) - \rho_F(\bar{\omega} - \bar{x}) = -T_Q^F(\bar{x})
\]
by \( -x^* \in \partial \rho_F(\bar{\omega} - \bar{x}) \). It follows therefore that \( \langle x^*, x - \bar{x} \rangle \leq \epsilon \) for all \( x \in Q_r \), and hence \( x^* \in N(\bar{x}; Q_r) \) because \( \epsilon > 0 \) was chosen arbitrary small. \( \square \)

Let \( \varphi : X \to (-\infty, \infty] \) be a convex function and let \( \bar{x} \in \text{dom} \varphi \). The singular subdifferential of \( \varphi \) at \( \bar{x} \) is defined by
\[
\partial^\infty \varphi(\bar{x}) = \{ x^* \in X^* : (x^*, 0) \in N((\bar{x}, \varphi(\bar{x})); \text{epi} \varphi) \}.
\]

**Lemma 4.3** Let \( \varphi : X \to (-\infty, \infty] \) be a convex function and let \( \bar{x} \in \text{dom} \varphi \). We always have
\[
\partial^\infty \varphi(\bar{x}) = N(\bar{x}; \text{dom} \varphi).
\]

**Proof.** Fix any \( x^* \in \partial^\infty \varphi(\bar{x}) \) and \( x \in \text{dom} \varphi \). Then \( (x, \varphi(x)) \in \text{epi} \varphi \). Thus
\[
\langle x^*, x - \bar{x} \rangle = \langle x^*, x - \bar{x} \rangle + 0(\varphi(x) - \varphi(\bar{x})) \leq 0
\]
because \( (x^*, 0) \in N((\bar{x}, \varphi(\bar{x})); \text{epi} \varphi) \). Conversely, suppose \( x^* \in N(\bar{x}; \text{dom} \varphi) \). Fix any \( (x, \lambda) \in \text{epi} f \). Then \( \varphi(x) \leq \lambda \), so \( x \in \text{dom} \varphi \). Thus
\[
\langle x^*, x - \bar{x} \rangle + 0(\lambda - \varphi(\bar{x})) = \langle x^*, x - \bar{x} \rangle \leq 0.
\]
This implies \( (x^*, 0) \in N((\bar{x}, \varphi(\bar{x})); \text{epi} \varphi) \) or equivalently \( x^* \in \partial^\infty \varphi(\bar{x}) \). \( \square \)

Define the following subset of \( X^* \)
\[
F_+^* = \{ x^* \in X^* : \langle x^*, f \rangle \geq 0 \text{ for all } f \in F \}.
\]
We are going to obtain below explicit representations of singular subdifferentials of the minimal time function (1).

**Proposition 4.1** Consider the minimal time function (1) with a nonempty convex target set \( Q \). The following hold:
(i) If \( \bar{x} \in Q \), then
\[
\partial^\infty \mathcal{T}_Q^F(\bar{x}) = N(\bar{x}; Q) \cap F_+^*.
\]
(ii) If \( \bar{x} \in Q - F_\infty \) and \( 0 \in F \), then
\[
\partial^\infty \mathcal{T}_Q^F(\bar{x}) = N(\bar{x}; Q - F_\infty) \cap F_+^*.
\]
(iii) If \( \bar{x} \notin Q - F_\infty \), \( X \) is a reflexive Banach space, and \( Q \) is closed and bounded, \( 0 \in F \), and \( \mathcal{T}_Q^F(\bar{x}) < \infty \), then
\[
\partial^\infty \mathcal{T}_Q^F(\bar{x}) = N(\bar{x}; Q_r) \cap F_+^*, \ r = \mathcal{T}_Q^F(\bar{x}) > 0.
\]
The smallest intersecting ball problem (SIB) under consideration asks for a point \( \bar{x} \). In this section, we consider the optimization problem (2), where the target sets \( \Omega \) satisfy \( \bar{x} \). The proof of (i) follows by the same proof of (i) using the observation that

\[
\langle x^*, f \rangle \geq 0.
\]

Let us prove the converse. Fix any \( x^* \in N(\bar{x}; Q) \) such that (i) holds. Since 0 is a solution of the smallest intersecting ball problem (SIB) with the smallest radius \( r \) if for all \( i = 1, \ldots, m \), \( x^* \in N(\bar{x}; \Omega_i) \). The proof for (iii) is similar.

5 Applications to the Smallest Intersecting Ball Problem

In this section, we consider the optimization problem (2), where the target sets \( \Omega_i \) for \( i = 1, \ldots, m \) and the constraint set \( \Omega \) are nonempty closed bounded convex subsets of \( X \). The smallest intersecting ball problem (SIB) under consideration asks for a point \( \bar{x} \in \Omega \) and the smallest \( t \geq 0 \) (if they exist) such that

\[
(x^* + tf) \subset \Omega_i \neq \emptyset \text{ for all } i = 1, \ldots, m.
\]

**Definition 5.1** We say that \( \bar{x} \in \Omega \) is a solution of the smallest intersecting ball problem (SIB) with the smallest radius \( r \) if \( r \in [0, \infty) \),

\[
(x^* + rF) \subset \Omega_i \neq \emptyset \text{ for all } i = 1, \ldots, m,
\]

and whenever there exist \( x \in \Omega \) and \( r' \in [0, \infty) \) with

\[
(x + r'F) \subset \Omega_i \neq \emptyset \text{ for all } i = 1, \ldots, m,
\]

then \( r \leq r' \).

**Lemma 5.1** Let \( X \) be a reflexive Banach space. Suppose \( Q \subseteq X \) is a nonempty closed bounded convex set and \( 0 \in F \). If \( T^F_Q(\bar{x}) \leq r \) and \( r > 0 \), then \( (\bar{x} + rF) \cap Q \neq \emptyset \).
Proof. Let \( \gamma = T_Q^F(\bar{x}) \). Suppose first that \( \gamma > 0 \). Then there exists a sequence \( \{t_k\} \) that converges to \( \gamma \) such that \( \bar{x} + t_k f_k = \omega_k \), where \( f_k \in F \) and \( \omega_k \in Q \). Since \( Q \) is closed convex and bounded, we can assume that \( \omega_k \xrightarrow{w} \omega \in Q \). Then \( f_k \xrightarrow{w} \frac{\omega - \bar{x}}{\gamma} \). Since \( F \) is closed and convex, it is weakly closed, and hence \( \omega - \bar{x} \in F \). Thus \( (\bar{x} + \gamma F) \cap Q \neq \emptyset \). It follows that \( (\bar{x} + r F) \cap Q \neq \emptyset \) since \( \gamma \leq r \) and \( 0 \in F \). If \( \gamma = 0 \), then \( \bar{x} \in Q - F \) by Theorem 3.1. Let \( \bar{x} = \omega - d \), where \( \omega \in Q \) and \( d \in F_\infty \). By Corollary 2.1, for \( r \geq 0 \)

\[
\frac{\omega - \bar{x}}{r} = \frac{d}{r} \in F.
\]

This again implies \( (\bar{x} + r F) \cap Q \neq \emptyset \). The proof is now complete.

\[
\square
\]

Theorem 5.1 Let \( X \) be a reflexive Banach space and let \( 0 \in \text{int} F \). Suppose \( \Omega \) and \( \Omega_i \) for \( i = 1, \ldots, m \) are nonempty closed bounded convex sets and

\[
\Omega \cap \bigcap_{i=1}^{m} (\Omega_i - F_\infty) = \emptyset. \tag{8}
\]

Then \( \bar{x} \in \Omega \) is an optimal solution of the optimization problem (2) with \( r = T(\bar{x}) \) if and only if \( \bar{x} \) is an optimal solution of the smallest intersecting ball problem (SIB) with the smallest radius \( r \).

Proof. Suppose \( \bar{x} \in \Omega \) is an optimal solution of (2) and \( r = T(\bar{x}) \). If \( r = 0 \), then \( T_{\Omega_i}^F(\bar{x}) = 0 \) for all \( i = 1, \ldots, m \). This implies \( \bar{x} \in \Omega_i - F_\infty \) for all \( i = 1, \ldots, m \), which is a contradiction to (8). Thus \( r > 0 \). From Lemma 5.1, \( (\bar{x} + r F) \cap \Omega_i \neq \emptyset \) for all \( i = 1, \ldots, m \). Let us fix an \( x \in \Omega \) and \( r' \geq 0 \) such that \( (x + r' F) \cap \Omega_i \neq \emptyset \) for all \( i = 1, \ldots, m \). We will show that \( r' \geq r \). Suppose \( r' < r \). Since \( T_{\Omega_i}^F(x) \leq r' \) for all \( i = 1, \ldots, m \), one has \( T(x) \leq r' < r = T(\bar{x}) \), which is a contradiction.

Let us prove the converse. Suppose that \( \bar{x} \) is an optimal solution of the smallest intersecting ball problem (SIB) with the smallest radius \( r \). From Definition 5.1, one has \( \bar{x} \in \Omega \) and \( (\bar{x} + r F) \cap \Omega_i \neq \emptyset \) for all \( i = 1, \ldots, m \). Thus

\[
T(\bar{x}) = \max \{ T_{\Omega_i}^F(\bar{x}) : i = 1, \ldots, m \} \leq r. \tag{9}
\]

By (8), \( r > 0 \). Fix any \( x \in \Omega \) and define \( r' = T(x) \). Since \( 0 \in \text{int} F \) and by (8), one has \( r' \in (0, \infty) \). By the definition of the minimal time function (1), \( T_{\Omega_i}^F(x) \leq r' \) for all \( i = 1, \ldots, m \). It follows again from Lemma 5.1 that

\[
(x + r' F) \cap \Omega_i \neq \emptyset \text{ for all } i = 1, \ldots, m.
\]

Thus \( r \leq r' = T(x) \) by Definition 5.1. In particular, taking \( x = \bar{x} \), one has \( r \leq T(\bar{x}) \), and hence \( r = T(\bar{x}) \) by (9). So \( T(\bar{x}) \leq T(x) \). Therefore, \( \bar{x} \) is an optimal solution of the optimization problem (2).

For each \( u \in X \), define

\[
I(u) = \{ i \in \{1, \ldots, m\} : T(u) = T_{\Omega_i}^F(u) \}.
\]
Theorem 5.2 Let $X = \mathbb{R}^n$ with the Euclidean norm and let $0 \in \text{int } F$. Consider the optimization problem (2) in which $\Omega_i$ for $i = 1, \ldots, m$ and $\Omega$ are nonempty closed bounded convex sets. Fix $x_1 \in \Omega$ and define the sequences of iterates by

$$x_{k+1} = \mathcal{P}(x_k - \alpha_k x^*_k; \Omega), \ \forall k \in \mathbb{N},$$

where

$$x^*_k \in \begin{cases} 
\{0\} & \text{if } x_k \in \Omega_i - F_\infty, \\
[-\partial \rho_F(\omega_k - x_k)] \cap N(\omega_k; \Omega_i) & \text{if } x_k \notin \Omega_i - F_\infty,
\end{cases}$$

for some $i \in I(x_k)$, and $\mathcal{P}$ denotes the Euclidean projection operator.

Let $V_k = \min \{ T(x_j) : j = 1, \ldots, k \}$. If $\sum_{i=1}^\infty \alpha_i = \infty$ and $\sum_{i=1}^\infty \alpha_i^2 < \infty$, then the sequence $\{x_k\}$ converges to an optimal solution of problem (2) and $V_k$ converges to the optimal value $\overline{V}$ of the problem. Moreover,

$$V_k - \overline{V} \leq \frac{d(x_1; S)^2 + L^2 \sum_{i=1}^k \alpha_i^2}{2 \sum_{i=1}^k \alpha_i},$$

where $0 \leq L < \infty$ is a Lipschitz constant of the function $T$, and $S$ is the solution set of problem (2).

Proof. First observe that $S \neq \emptyset$ since $T$ is a continuous function and the constraint $\Omega$ is compact. We have the following well-known formula, see e.g. [9, Corollary 4.3.2], for computing the subdifferential of the function $T$ at a point $u$:

$$\partial T(u) = \text{co} \{ \partial T^F_{\Omega_i}(u) : i \in I(u) \}.$$ 

Thus for any $i \in I(u)$, one has $\partial T^F_{\Omega_i}(u) \subset \partial T(u)$. The choice of $x^*_k$ in the statement of the theorem allows us to find an element of $\partial T^F_{\Omega_i}(x_k)$ by Theorem 4.1, Theorem 4.2, and Theorem 4.4. Then $x^*_k \in \partial T(x_k)$. The rest of the proof comes from the standard projective subgradient method; see [1].

In the proposition below, we provide explicit formulas for computing the asymptotic cone and the Minkowski function generated by an unbounded dynamic.

Proposition 5.1 Let $x^*_i \in X^*$ and $b_i > 0$ for $i = 1, \ldots, q$. Define

$$F = \{ x \in X : \langle x^*_i, x \rangle \leq b_i \text{ for all } i = 1, \ldots, q \}. \quad (10)$$

Then

$$F_\infty = \{ x \in X : \langle x^*_i, x \rangle \leq 0 \text{ for all } i = 1, \ldots, q \}$$

and

$$\rho_F(u) = \max \{ 0, \max \{ \frac{\langle x^*_i, u \rangle}{b_i} : i = 1, \ldots, q \} \}.$$
**Proof.** Let \( x^* \in X^* \). By the definition of \( F \) we have that \( F \) is a closed convex set. Since \( 0 \in F \), by Corollary 2.1, one has that \( d \in F_\infty \) if and only if
\[
td \in F \quad \text{for all } t > 0,
\]
or
\[
\langle x_i^*, d \rangle \leq \frac{b_i}{t} \quad \text{for all } i = 1, \ldots, q \text{ and for all } t > 0.
\]
This is equivalent to
\[
\langle x_i^*, d \rangle \leq 0 \quad \text{for all } i = 1, \ldots, q.
\]
The formula for \( F_\infty \) has been verified.

Since \( b_i > 0 \) for all \( i = 1, \ldots, q \), one has \( 0 \in \text{int } F \), and
\[
\rho_F(u) = \inf \{ t > 0 : u \in tF \}
\]
\[
= \inf \{ t > 0 : \langle x_i^*, u \rangle \leq t b_i \text{ for all } i = 1, \ldots, q \}
\]
\[
= \inf \{ t > 0 : \frac{\langle x_i^*, u \rangle}{b_i} \leq t \text{ for all } i = 1, \ldots, q \}
\]
\[
= \max \{ 0, \max \{ \frac{\langle x_i^*, u \rangle}{b_i} : i = 1, \ldots, q \} \}.
\]
The proof is now complete. \( \square \)

The following corollary provides a formula for computing the subdifferential of \( \rho_F(u) \) when \( u \notin F_\infty \). The formula allows us to determine a subgradient \( x_k^* \) in Theorem 5.2 when \( F \) is the polyhedron dynamic (10) in the case \( x_k \notin \Omega_i - F_\infty \).

**Corollary 5.1** Consider the set \( F \) defined in Proposition 5.1. For \( u \in X \), define
\[
J(u) = \{ i = 1, \ldots, p : \frac{\langle x_i^*, u \rangle}{b_i} = \rho_F(u) \}.
\]
For any \( u \notin F_\infty \), one has \( J(u) \neq \emptyset \) and
\[
\partial \rho_F(u) = \text{co} \{ \frac{x_i^*}{b_i} : i \in J(u) \}.
\]

**Proof.** Since \( u \notin F_\infty \), by Proposition 5.1, there exists \( j = 1, \ldots, q \) such that \( \langle x_j^*, u \rangle > 0 \). This implies \( \rho_F(u) > 0 \) and \( J(u) \neq \emptyset \). Using the formula for computing \( \rho_F \) in Proposition 5.1 and the well-known formula for computing subdifferential of “max” function, see e.g. [9, Corollary 4.3.2] (which also holds in this setting), one obtains (11). \( \square \)

The following example shows an application of the subgradient method.

**Example 5.1** Let \( F = \{ (x, y) \in \mathbb{R}^2 : x \leq 1, y \leq 1 \} \) where \( x_1^* = (1,0) \) and \( x_2^* = (0,1) \). Then \( F_\infty = \{ (x, y) : x \leq 0, y \leq 0 \} \). Define the target sets \( \Omega_i \) by two-tuples, that is, \( \Omega_i = \{ (a_i, b_i) \} \) for \( i = 1, \ldots, m \) and let \( \Omega = \mathbb{B}(c; r) \). Then \( \rho_F((u_1, u_2)) = \max \{ u_1, u_2, 0 \} \), and \( \mathcal{T}_{\Omega_i}^F((u_1, u_2)) = \rho_F((a_i, b_i) - (u_1, u_2)) = \max \{ 0, a_i - u_1, b_i - u_2 \} \). Suppose that we have found \( x_k = (x_{1k}, x_{2k}) \) from the subgradient algorithm. Then we compute \( \mathcal{T}_{\Omega_i}^F(x_k) = \max \{ 0, a_i - x_{1k}, b_i - x_{2k} \} \) and \( \mathcal{T}(x_k) = \max \{ \mathcal{T}_{\Omega_i}^F(x_k) : i = 1, \ldots, m \} \). Fix \( i \in I(x_k) \). If
\(x_{1k} \geq a_i\) and \(x_{2k} \geq b_i\), then choose \(x_k^* = 0\). Otherwise, \(x_k \in \Omega_i - F_{\infty}\). If \(a_i - x_{1k} \geq b_i - x_{2k}\), then choose \(x_k^* = (-1,0)\). In the case \(a_i - x_{1k} < b_i - x_{2k}\), choose \(x_k^* = (0,-1)\). The Euclidean projection to \(\Omega\) can be easily obtained. Thus we can determine \(x_{k+1}\), and the algorithm is now explicit.

| MATLAB RESULT |
|----------------|
| \(k\) | \(x_k\) | \(V_k\) |
| 1 | (-7.00000, -4.00000) | 13.00000 |
| 10 | (-4.17103, -4.00000) | 10.28214 |
| 100 | (-3.10201, -3.10240) | 9.10252 |
| 1,000 | (-3.10210, -3.10200) | 9.10216 |
| 10,000 | (-3.10210, -3.10201) | 9.10211 |
| 100,000 | (-3.10209, -3.10208) | 9.10209 |
| 200,000 | (-3.10208, -3.10208) | 9.10208 |
| 300,000 | (-3.10208, -3.10208) | 9.10208 |
| 400,000 | (-3.10208, -3.10208) | 9.10208 |
| 500,000 | (-3.10208, -3.10208) | 9.10208 |

Figure 1: A smallest intersecting ball problem to squares with an unbounded dynamic.

**Example 5.2** Let \(F\) be the same set as in the previous example. Let \(\Omega_i\) be a square of right position \([\omega_{1i} - r_i, \omega_{1i} + r_i] \times [\omega_{2i} - r_i, \omega_{2i} + r_i]\) with center \(c_i = (\omega_{1i}, \omega_{2i})\) and short radius \(r_i\) for \(i = 1, \ldots, m\). Define the constraint \(\Omega\) to be a nonempty closed bounded convex set. In this case,

\[
\Omega_i - F_{\infty} = \{(x_1, x_2) : x_1 \geq \omega_{1i} - r_i, x_2 \geq \omega_{2i} - r_i\}
\]

and the projection \(\tilde{\omega}\) can be represented by \(c_i\) and \(r_i\).

The sequence \(\{x_k^*\}\) in Theorem 5.2 can be determined as follows

\[
x_k^* = \begin{cases} 
(-1,0), & \text{if } x_{1k} < \omega_{1i} - r_i \text{ and } x_{2k} - \omega_{2i} \geq x_{1k} - \omega_{1i}, \\
(0,-1), & \text{if } x_{2k} < \omega_{2i} - r_i \text{ and } x_{2k} - \omega_{2i} < x_{1k} - \omega_{1i}, \\
(0,0), & \text{otherwise}
\end{cases}
\]

where \(i \in I(x_k)\). The sequence \(\{V_k\}\) can be determined based on the fact that \(i \in I(x_k)\), so we have

\[
\mathcal{T}^F_{\Omega_i}((x_{1k}, x_{2k})) = \begin{cases} 
(\omega_{1i} - r_i) - x_{1k}, & \text{if } x_{1k} < \omega_{1i} - r_i \text{ and } x_{2k} - \omega_{2i} \geq x_{1k} - \omega_{1i}, \\
(\omega_{2i} - r_i) - x_{2k}, & \text{if } x_{2k} < \omega_{2i} - r_i \text{ and } x_{2k} - \omega_{2i} < x_{1k} - \omega_{1i}, \\
0, & \text{otherwise}.
\end{cases}
\]

Consider the target sets \(\Omega_i, i = 1, 2, \ldots, 8\), to be the right squares with centers \((-30,0), (-26,-34), (-20,-15), (-13,8), (-9,-25), (2,-1), (8,-30)\) and \((8,5)\) with radii \(r_i = 2\) for all \(i = 1, 2, \ldots, 8\), and the constraint set \(\Omega\) is the disk with center \((-7,-4)\) and radius 4. Choose \(\alpha_k = 1/k\) that satisfies Theorem 5.2 and the starting point \((-7,-4)\). An optimal solution and the optimal value are \(\bar{x} \approx (-3.10208, -3.10208)\) and \(\bar{V} \approx 9.10208\); see Figure 1.
6 Concluding Remarks

This paper is a part of our project involving *set facility location problems*. The main idea is to consider a much broader situation where singletons in the classical models of facility location problems are replaced by sets. After developing new results for general and generalized differentiation properties of minimal time functions with unbounded dynamics, we are able to provide new applications to the smallest intersecting ball problem with convex target sets.

Our next goal is to study the smallest intersecting ball problem with unbounded dynamics and nonconvex target sets. To achieve this goal, we need to develop nonconvex generalized differentiation theory for minimal time functions generated by unbounded dynamics. We foresee the potential of success of this future work.

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