

CONVERGENCE OF THE PATH AND ITS DISCRETIZATION TO THE MINIMUM-NORM FIXED POINT OF PSEUDOCONTRACTIONS

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ABSTRACT. Let C be a nonempty closed convex subset of a real Hilbert space H . Let $T : C \rightarrow C$ be a Lipschitz pseudocontractive mapping with $Fix(T) \neq \emptyset$. In this paper, we first show that as $t \rightarrow 0+$, the path $x \rightarrow x_t, t \in (0, 1)$, in C , defined by $x_t = (1 - \beta)P_C[(1 - t)x_t] + \beta Tx_t$ converges strongly to the minimum-norm fixed point of T . Subsequently, by discretizing the path, we suggest an explicit method $x_{n+1} = (1 - \beta_n)P_C[(1 - \alpha_n)x_n] + \beta_n Tx_n$. Under some assumptions, we prove the sequence $\{x_n\}$ also converges strongly to the minimum-norm fixed point of T .

1. INTRODUCTION

The interest of pseudocontractions lies in their connection with monotone operators; namely, T is a pseudocontraction if and only if the complement $I - T$ is a monotone operator. However, it is now well-known that Mann's algorithm fails to converge for Lipschitzian pseudocontractions (see the counterexample of Chidume and Mutangadura [1]). It is therefore an interesting question of inventing iterative algorithms which generate a sequence converging in the norm topology to a fixed point of a Lipschitzian pseudocontraction (if any). On the other hand, it is quite often to seek a particular solution of a given nonlinear problem, in particular, the minimum-norm solution.

Recently, in order to find the minimum-norm fixed point of Lipschitzian pseudocontractions, Yao, Colao, Marino and Xu [4] suggested the following implicit algorithm

$$(1.1) \quad x_n = \beta_n x_{n-1} + (1 - \alpha_n - \beta_n)Tx_n, n \geq 0.$$

They proved that under some mild assumptions on algorithm parameters $\{\alpha_n\}$ and $\{\beta_n\}$, the sequence $\{x_n\}$ defined by (1.1) converges strongly to the minimum-norm fixed point of T provided $0 \in C$. They pointed out that this assumption $0 \in C$ cannot be removed due to the algorithm (1.1) may not be well-defined. Afterwards, they further put forth the following interesting topic: It is of interest to adapt the algorithm (1.1) to suit for the general case (i.e., without assuming $0 \in C$) of find the minimum-norm fixed point of a Lipschitz pseudocontraction.

2010 *Mathematics Subject Classification.* 47H05, 47H10, 47H17.

Key words and phrases. Pseudocontraction, implicit algorithm, explicit method, minimum-norm, fixed point, projection.

Yonghong Yao was supported in part by NSFC 71161001-G0105. Rudong Chen was supported in part by NSFC 11071279. Yeong-Cheng Liou was supported in part by NSC 101-2628-E-230-001-MY3 and NSC 101-2622-E-230-005-CC3.

The purpose of this paper is to construct an implicit algorithm which defines a net $\{x_t\}$ converging strongly to the minimum-norm fixed point of a Lipschitz pseudocontraction without assuming $0 \in C$. Subsequently, by discretizing the net, we suggest an explicit method which generates a sequence $\{x_n\}$. Under some assumptions, we prove the sequence $\{x_n\}$ also converges strongly to the minimum-norm fixed point of T .

2. PRELIMINARIES

Let H be a real Hilbert space with inner product $\langle \cdot, \cdot \rangle$ and norm $\|\cdot\|$, respectively. Let C be a nonempty closed convex subset of H . Recall the following notions for a mapping $T : C \rightarrow C$.

- T is called pseudocontractive (or a pseudocontraction) if

$$\langle Tx - Ty, x - y \rangle \leq \|x - y\|^2, \quad x, y \in C;$$

- T is nonexpansive if

$$\|Tx - Ty\| \leq \|x - y\|, \quad \forall x, y \in C.$$

The (nearest point or metric) projection from H onto C is defined as follows: for each point $x \in H$, $P_C x$ is the unique point in C with the property:

$$\|x - P_C x\| \leq \|x - y\|, \quad y \in C.$$

Note that P_C is characterized by the inequality:

$$P_C x \in C, \quad \langle x - P_C x, y - P_C x \rangle \leq 0, \quad y \in C.$$

We adopt the following notations:

- $Fix(T)$ stands for the set of fixed points of T ;
- $x_n \rightharpoonup x$ stands for the weak convergence of $\{x_n\}$ to x ;
- $x_n \rightarrow x$ stands for the strong convergence of $\{x_n\}$ to x .

We need the following lemma for proof of our main results.

Lemma 2.1 ([5]). *Let C be a closed convex subset of a Hilbert space H . Let $T : C \rightarrow C$ be a Lipschitz pseudocontraction. Then $Fix(T)$ is a closed convex subset of C and the mapping $I - T$ is demiclosed at 0, i.e. whenever $\{x_n\} \subset C$ is such that $x_n \rightharpoonup x$ and $(I - T)x_n \rightarrow 0$, then $(I - T)x = 0$.*

Lemma 2.2 ([3]). *Assume $\{a_n\}$ is a sequence of nonnegative real numbers such that*

$$a_{n+1} \leq (1 - \gamma_n)a_n + \gamma_n \delta_n, \quad n \geq 0,$$

where $\{\gamma_n\}$ is a sequence in $(0, 1)$ and $\{\delta_n\}$ is a sequence in \mathbb{R} such that

- (i) $\sum_{n=0}^{\infty} \gamma_n = \infty$;
- (ii) $\limsup_{n \rightarrow \infty} \delta_n \leq 0$ or $\sum_{n=0}^{\infty} |\delta_n \gamma_n| < \infty$.

Then $\lim_{n \rightarrow \infty} a_n = 0$.

3. MAIN RESULTS

In this section, we introduce our algorithms and prove the strong convergence of these algorithms to the minimum norm fixed point of pseudocontractive mapping T .

First, we introduce an implicit path on pseudocontractive mappings.

Algorithm 3.1. *Let C be a nonempty closed convex subset of a real Hilbert space H . Let $T : C \rightarrow C$ be Lipschitz a pseudocontraction. Let $\beta \in (0, 1)$ be a constant. For each $t \in (0, 1)$, let the net $\{x_t\}$ be defined as the unique solution of fixed point equation*

$$(3.1) \quad x_t = (1 - \beta)P_C[(1 - t)x_t] + \beta Tx_t, \quad t \in (0, 1),$$

where $P_C : H \rightarrow C$ is the metric projection from H on C .

Remark 3.2. We note that the algorithm (3.1) is well-defined. Indeed, for $\beta, t \in (0, 1)$ define a mapping $U_t : C \rightarrow C$ by

$$U_t x = (1 - \beta)P_C[(1 - t)x] + \beta Tx, \quad x \in C.$$

It is clear that U_t is a self-mapping of C . For $x, y \in C$, we have

$$\begin{aligned} \langle U_t x - U_t y, x - y \rangle &= (1 - \beta) \langle P_C[(1 - t)x] - P_C[(1 - t)y], x - y \rangle \\ &\quad + \beta \langle Tx - Ty, x - y \rangle \\ &\leq (1 - \beta) \|P_C[(1 - t)x] - P_C[(1 - t)y]\| \|x - y\| \\ &\quad + \beta \|x - y\|^2 \\ &\leq (1 - \beta)(1 - t) \|x - y\|^2 + \beta \|x - y\|^2 \\ &= [1 - (1 - \beta)t] \|x - y\|^2. \end{aligned}$$

This implies that U_t is strongly pseudocontractive. So, by Deimling [2], U_t has a unique fixed point $x_t \in C$ which is the unique solution of the fixed point equation (3.1).

We are now in a position to prove the strong convergence of the implicit algorithm (3.1) to the minimum-norm fixed point of the pseudocontractive mapping T .

Theorem 3.3. *Let C be a nonempty closed convex subset of a real Hilbert space H . Let $T : C \rightarrow C$ be a Lipschitz pseudocontraction with $\text{Fix}(T) \neq \emptyset$. Then the net $\{x_t\}$ defined by (3.1) converges in norm, as $t \rightarrow 0^+$, to the minimum-norm fixed point of T .*

Proof. We first show that the net $\{x_t\}$ is bounded.

Taking $p \in \text{Fix}(T)$, we get from (3.1) that

$$\begin{aligned} \|x_t - p\|^2 &= (1 - \beta) \langle P_C[(1 - t)x_t] - p, x_t - p \rangle + \beta \langle Tx_t - p, x_t - p \rangle \\ &\leq (1 - \beta) \|P_C[(1 - t)x_t] - p\| \|x_t - p\| + \beta \|x_t - p\|^2 \\ &\leq (1 - \beta) \|(1 - t)x_t - p\| \|x_t - p\| + \beta \|x_t - p\|^2 \\ &\leq (1 - \beta) [(1 - t) \|x_t - p\| + t \|p\|] \|x_t - p\| + \beta \|x_t - p\|^2. \end{aligned}$$

It turns that

$$\|x_t - p\| \leq \|p\|.$$

Consequently, $\{x_t\}$ is bounded and so is $\{Tx_t\}$.

Next, we show that $\lim_{t \rightarrow 0^+} \|x_t - Tx_t\| = 0$.

From (3.1), we have

$$\begin{aligned} \|x_t - Tx_t\| &= \|(1 - \beta)P_C[(1 - t)x_t] + \beta Tx_t - Tx_t\| \\ &\leq (1 - \beta)\|P_C[(1 - t)x_t] - Tx_t\| \\ &\leq (1 - \beta)(\|x_t - Tx_t\| + t\|x_t\|). \end{aligned}$$

Therefore,

$$(3.2) \quad \|x_t - Tx_t\| \leq \frac{(1 - \beta)t}{\beta} \|x_t\| \rightarrow 0.$$

Next we show that $\{x_t\}$ is relatively norm-compact as $t \rightarrow 0^+$. Let $\{t_n\} \subset (0, 1)$ be a sequence such that $t_n \rightarrow 0^+$ as $n \rightarrow \infty$. Put $x_n := x_{t_n}$. From (3.2), we have

$$(3.3) \quad \|x_n - Tx_n\| \rightarrow 0.$$

Again from (3.1), we get

$$\begin{aligned} \|x_t - p\|^2 &= (1 - \beta)\langle P_C[(1 - t)x_t] - p, x_t - p \rangle + \beta\langle Tx_t - p, x_t - p \rangle \\ &\leq (1 - \beta)\|P_C[(1 - t)x_t] - p\|\|x_t - p\| + \beta\|x_t - p\|^2 \\ &\leq (1 - \beta)\frac{1}{2}(\|P_C[(1 - t)x_t] - p\|^2 + \|x_t - p\|^2) + \beta\|x_t - p\|^2. \end{aligned}$$

It follows that

$$\begin{aligned} \|x_t - p\|^2 &\leq \|P_C[(1 - t)x_t] - p\|^2 \\ &\leq \|x_t - p - tx_t\|^2 \\ &= \|x_t - p\|^2 - 2t\langle x_t, x_t - p \rangle + t^2\|x_t\|^2 \\ &= \|x_t - p\|^2 - 2t\langle x_t - p, x_t - p \rangle - 2t\langle p, x_t - p \rangle + t^2\|x_t\|^2 \\ &= (1 - 2t)\|x_t - p\|^2 - 2t\langle p, x_t - p \rangle + t^2\|x_t\|^2. \end{aligned}$$

It turns out that

$$(3.4) \quad \begin{aligned} \|x_t - p\|^2 &\leq \langle p, p - x_t \rangle + \frac{t}{2}\|x_t\|^2 \\ &\leq \langle p, p - x_t \rangle + tM. \end{aligned}$$

where $M > 0$ is some constant such that $\sup\{\frac{1}{2}\|x_t\|^2 : t \in (0, 1)\} \leq M$. In particular, we get from (3.4)

$$(3.5) \quad \|x_n - p\|^2 \leq \langle p, p - x_n \rangle + t_n M, \quad p \in \text{Fix}(T).$$

Since $\{x_n\}$ is bounded, without loss of generality, we may assume that $\{x_n\}$ converges weakly to a point $x^* \in C$. Noticing (3.3) we can use Lemma 2.1 to get $x^* \in \text{Fix}(T)$. Therefore we can substitute x^* for p in (3.5) to get

$$(3.6) \quad \|x_n - x^*\|^2 \leq \langle x^*, x^* - x_n \rangle + t_n M.$$

However, $x_n \rightharpoonup x^*$. This together with (3.6) guarantees that $x_n \rightarrow x^*$. The net $\{x_t\}$ is therefore relatively compact, as $t \rightarrow 0^+$, in the norm topology.

Now we return to (3.5) and take the limit as $n \rightarrow \infty$ to get

$$\|x^* - p\|^2 \leq \langle p, p - x^* \rangle, \quad p \in \text{Fix}(T).$$

This is equivalent to

$$(3.7) \quad 0 \leq \langle x^*, p - x^* \rangle, \quad p \in \text{Fix}(T).$$

Therefore, $x^* = P_{\text{Fix}(T)}0$. This is sufficient to conclude that the entire net $\{x_t\}$ converges in norm to x^* and x^* is the minimum-norm fixed point of T . As a matter of fact, from (3.7), we have

$$\|x^*\|^2 \leq \langle x^*, p \rangle \leq \|x^*\| \|p\|, \quad p \in \text{Fix}(T).$$

It follows that

$$\|x^*\| \leq \|p\|, \quad p \in \text{Fix}(T).$$

This completes the proof. \square

Now, we introduce an explicit algorithm which is the discretization of (3.1) and prove its strong convergence to the minimum-norm fixed point of T .

Algorithm 3.4. *Let C be a nonempty closed convex subset of a real Hilbert space H . Let $T : C \rightarrow C$ be a pseudocontraction. Let $\{\alpha_n\}$ and $\{\beta_n\}$ be two real number sequences in $(0, 1)$. For chosen $x_0 \in C$ arbitrarily, we define a sequence $\{x_n\}$ iteratively by the following manner*

$$(3.8) \quad x_{n+1} = (1 - \beta_n)P_C[(1 - \alpha_n)x_n] + \beta_n T x_n, \quad n \geq 0.$$

Theorem 3.5. *Let C be a nonempty closed convex subset of a real Hilbert space H , and let $T : C \rightarrow C$ be an L -Lipschitzian pseudocontraction with $\text{Fix}(T) \neq \emptyset$. If the sequences $\{\alpha_n\}$ and $\{\beta_n\}$ satisfy the condition $\lim_{n \rightarrow \infty} \alpha_n = \lim_{n \rightarrow \infty} \frac{\alpha_n}{\beta_n} = \lim_{n \rightarrow \infty} \frac{\beta_n^2}{\alpha_n} = 0$, then the following hold:*

- (i) *the sequence $\{x_n\}$ defined by (3.8) is bounded;*
- (ii) *the sequence $\{x_n\}$ is asymptotically regular, that is, $\lim_{n \rightarrow \infty} \|x_{n+1} - x_n\| = 0$.*

Further, if $\sum_{n=0}^{\infty} \alpha_n = \infty$ and $\lim_{n \rightarrow \infty} \frac{\|x_{n+1} - x_n\|}{\beta_n} = 0$, then the sequence $\{x_n\}$ converges strongly to the minimum-norm fixed point of T .

Proof. First we prove that the sequence $\{x_n\}$ is bounded. We will show this fact by induction. According to the assumption, there exists a sufficiently large positive integer m such that

$$(3.9) \quad 1 - \frac{1}{1/2 - \beta_m} (L + 1)(L + 2) (\alpha_m + 2\beta_m + (\beta_m^2/\alpha_m)) > 0, \quad n \geq m.$$

Fix a $p \in \text{Fix}(T)$ and take a constant $M_1 > 0$ such that

$$(3.10) \quad \max\{\|x_0 - p\|, \|x_1 - p\|, \dots, \|x_{m-1} - p\|, 4\|x_m - p\| + 8\|p\|\} \leq M_1.$$

Next, we show that $\|x_{m+1} - p\| \leq M_1$. Set $y_m = P_C[(1 - \alpha_m)x_m]$, thus $x_{m+1} = (1 - \beta_m)y_m + \beta_m T x_m$. By the fact that $I - T$ is monotone, we have

$$(3.11) \quad \langle (I - T)x_{m+1} - (I - T)p, x_{m+1} - p \rangle \geq 0.$$

From (3.8), we obtain

$$\begin{aligned}
\|x_{m+1} - p\|^2 &= (1 - \beta_m)\langle y_m - p, x_{m+1} - p \rangle + \beta_m\langle Tx_m - p, x_{m+1} - p \rangle \\
&= (1 - \beta_m)\langle y_m - (1 - \alpha_m)x_m, x_{m+1} - p \rangle \\
&\quad + (1 - \beta_m)\langle (1 - \alpha_m)x_m - p, x_{m+1} - p \rangle \\
&\quad + \beta_m\langle Tx_m - p, x_{m+1} - p \rangle \\
&= (1 - \beta_m)\langle y_m - (1 - \alpha_m)x_m, x_{m+1} - p \rangle \\
&\quad + (1 - \beta_m)\langle x_m - p, x_{m+1} - p \rangle - (1 - \beta_m)\alpha_m\langle x_m, x_{m+1} - p \rangle \\
&\quad + \beta_m\langle Tx_m - p, x_{m+1} - p \rangle \\
&= (1 - \beta_m)\langle y_m - (1 - \alpha_m)x_m, x_{m+1} - p \rangle \\
&\quad + \langle x_m - p, x_{m+1} - p \rangle - (1 - \beta_m)\alpha_m\langle x_{m+1} - p, x_{m+1} - p \rangle \\
&\quad - (1 - \beta_m)\alpha_m\langle x_m - x_{m+1}, x_{m+1} - p \rangle \\
&\quad - (1 - \beta_m)\alpha_m\langle p, x_{m+1} - p \rangle + \beta_m\langle Tx_m - Tx_{m+1}, x_{m+1} - p \rangle \\
&\quad + \beta_m\langle x_{m+1} - x_m, x_{m+1} - p \rangle - \beta_m\langle x_{m+1} - Tx_{m+1}, x_{m+1} - p \rangle.
\end{aligned}$$

Then, from (3.11), we get

$$\begin{aligned}
\|x_{m+1} - p\|^2 &\leq (1 - \beta_m)\|y_m - (1 - \alpha_m)x_m\|\|x_{m+1} - p\| + \|x_m - p\|\|x_{m+1} - p\| \\
&\quad - (1 - \beta_m)\alpha_m\|x_{m+1} - p\|^2 + (1 - \beta_m)\alpha_m\|p\|\|x_{m+1} - p\| \\
&\quad + (1 - \beta_m)\alpha_m(\|x_{m+1} - x_m\| + \|p\|)\|x_{m+1} - p\| \\
&\quad + \beta_m(\|Tx_m - Tx_{m+1}\| + \|x_{m+1} - x_m\|)\|x_{m+1} - p\| \\
&\leq \|x_m - p\|\|x_{m+1} - p\| + 2(1 - \beta_m)\alpha_m(\|x_m - p\| + \|p\|)\|x_{m+1} - p\| \\
&\quad - (1 - \beta_m)\alpha_m\|x_{m+1} - p\|^2 + (1 - \beta_m)\alpha_m\|p\|\|x_{m+1} - p\| \\
&\quad + (1 - \beta_m)\alpha_m(\|x_{m+1} - x_m\| + \|p\|)\|x_{m+1} - p\| \\
&\quad + \beta_m(L + 1)\|x_{m+1} - x_m\|\|x_{m+1} - p\| \\
&\leq \|x_m - p\|\|x_{m+1} - p\| + (1 - \beta_m)\alpha_m(2\|x_m - p\| + 4\|p\|)\|x_{m+1} - p\| \\
&\quad - (1 - \beta_m)\alpha_m\|x_{m+1} - p\|^2 \\
&\quad + (\alpha_m + \beta_m)(L + 1)\|x_{m+1} - x_m\|\|x_{m+1} - p\|.
\end{aligned}$$

It follows that

$$\begin{aligned}
[1 + (1 - \beta_m)\alpha_m]\|x_{m+1} - p\| &\leq \|x_m - p\| + \alpha_m(2\|x_m - p\| + 4\|p\|) \\
(3.12) \qquad \qquad \qquad &\quad + (L + 1)(\alpha_m + \beta_m)\|x_{m+1} - x_m\|.
\end{aligned}$$

By (3.8), we have

$$\begin{aligned}
\|x_{m+1} - x_m\| &\leq (1 - \beta_m)\|P_C[(1 - \alpha_m)x_m] - P_C[x_m]\| + \beta_m\|Tx_m - x_m\| \\
&\leq (1 - \beta_m)\alpha_m(\|x_m - p\| + \|p\|) + \beta_m(\|Tx_m - p\| + \|p - x_m\|) \\
&\leq \alpha_m(\|x_m - p\| + \|p\|) + \beta_m(L + 1)\|x_m - p\| \\
&\leq (L + 1)(\alpha_m + \beta_m)\|x_m - p\| + \alpha_m\|p\| \\
(3.13) \qquad \qquad \qquad &\leq (L + 2)(\alpha_m + \beta_m)M_1.
\end{aligned}$$

Substitute (3.13) into (3.12) to obtain

$$\begin{aligned} & [1 + (1 - \beta_m)\alpha_m]\|x_{m+1} - p\| \\ & \leq \|x_m - p\| + \alpha_m(2\|x_m - p\| + 4\|p\|) + (L + 1)(L + 2)(\alpha_m + \beta_m)^2 M_1 \\ & \leq (1 + \frac{1}{2}\alpha_m)M_1 + (L + 1)(L + 2)(\alpha_m + \beta_m)^2 M_1. \end{aligned}$$

This together with (3.9) and (3.10) imply that

$$\begin{aligned} \|x_{m+1} - p\| & \leq \left[1 - \frac{(1/2 - \beta_m)\alpha_m - (L + 1)(L + 2)(\alpha_m + \beta_m)^2}{1 + (1 - \beta_m)\alpha_m} \right] M_1 \\ & = \left\{ 1 - \frac{(1/2 - \beta_m)\alpha_m [1 - \frac{1}{1/2 - \beta_m}(L + 1)(L + 2)(\alpha_m + 2\beta_m + (\beta_m^2/\alpha_m))]}{1 + (1 - \beta_m)\alpha_m} \right\} M_1 \\ & \leq M_1. \end{aligned}$$

By induction, we get

$$\|x_n - p\| \leq M_1, \quad \forall n \geq 0,$$

which implies that $\{x_n\}$ is bounded and so is $\{Tx_n\}$.

By (3.8), we have

$$\begin{aligned} \|x_n - Tx_n\| & \leq \|x_n - x_{n+1}\| + \|x_{n+1} - Tx_n\| \\ & \leq \|x_n - x_{n+1}\| + (1 - \beta_n)\|P_C[(1 - \alpha_n)x_n] - Tx_n\| \\ & \leq \|x_n - x_{n+1}\| + (1 - \beta_n)\|x_n - Tx_n\| + \alpha_n\|x_n\|. \end{aligned}$$

It follows that

$$\begin{aligned} \|x_n - Tx_n\| & \leq \frac{1}{\beta_n}\|x_n - x_{n+1}\| + \frac{\alpha_n}{\beta_n}\|x_n\| \\ & \leq \frac{1}{\beta_n}\|x_n - x_{n+1}\| + \frac{\alpha_n}{\beta_n}\|x_n\|. \end{aligned}$$

By the assumptions, we have

$$(3.14) \quad \lim_{n \rightarrow \infty} \|x_n - Tx_n\| = 0.$$

Next, we prove that

$$\limsup_{n \rightarrow \infty} \langle x^*, x^* - y_n \rangle \leq 0.$$

where $x^* = \lim_{t \rightarrow 0} z_t$ and $\{z_t\}$ is a net defined by $z_t = (1 - \beta)P_C[(1 - t)z_t] + \beta Tz_t$.

From the definition of $\{z_t\}$, we obtain

$$z_t - x_n = (1 - \beta)(P_C[(1 - t)z_t] - x_n) + \beta(Tz_t - Tx_n) + \beta(Tx_n - x_n).$$

It follows that

$$\begin{aligned} \|z_t - x_n\|^2 & = (1 - \beta)\langle P_C[(1 - t)z_t] - x_n, z_t - x_n \rangle + \beta\langle Tz_t - Tx_n, z_t - x_n \rangle \\ & \quad + \beta\langle Tx_n - x_n, z_t - x_n \rangle \\ & = (1 - \beta)\langle P_C[(1 - t)z_t] - (1 - t)z_t, z_t - x_n \rangle \\ & \quad + (1 - \beta)\langle (1 - t)z_t - x_n, z_t - x_n \rangle + \beta\langle Tz_t - Tx_n, z_t - x_n \rangle \\ & \quad + \beta\langle Tx_n - x_n, z_t - x_n \rangle. \end{aligned}$$

Noting that $x_n \in C$ and by using the property of the metric projection P_C , we have

$$\langle P_C[(1-t)z_t] - (1-t)z_t, z_t - x_n \rangle \leq 0.$$

So,

$$\begin{aligned} \|z_t - x_n\|^2 &\leq (1-\beta)\langle (1-t)z_t - x_n, z_t - x_n \rangle + \beta\|z_t - x_n\|^2 \\ &\quad + \beta\|Tx_n - x_n\|\|z_t - x_n\| \\ &= (1-\beta)\|z_t - x_n\|^2 - (1-\beta)t\langle z_t, z_t - x_n \rangle + \beta\|z_t - x_n\|^2 \\ &\quad + \beta\|Tx_n - x_n\|\|z_t - x_n\|. \end{aligned}$$

It follows that

$$\langle z_t, z_t - x_n \rangle \leq \frac{\beta}{(1-\beta)t}\|Tx_n - x_n\|\|z_t - x_n\|.$$

By (3.14), we deduce

$$(3.15) \quad \limsup_{t \rightarrow 0} \limsup_{n \rightarrow \infty} \langle z_t, z_t - x_n \rangle \leq 0.$$

Note the fact that the two limits $\limsup_{t \rightarrow 0}$ and $\limsup_{n \rightarrow \infty}$ are interchangeable. As a matter of fact, we have

$$\begin{aligned} \langle x^*, x^* - x_n \rangle &= \langle x^*, x^* - z_t \rangle + \langle x^* - z_t, z_t - x_n \rangle + \langle z_t, z_t - x_n \rangle \\ &\leq \langle x^*, x^* - z_t \rangle + \|x^* - z_t\|\|z_t - x_n\| + \langle z_t, z_t - x_n \rangle \\ &\leq (\|x^*\| + \|z_t - x_n\|)\|x^* - z_t\| + \langle z_t, z_t - x_n \rangle. \end{aligned}$$

This together with $z_t \rightarrow x^*$ and (3.15) imply that

$$\limsup_{n \rightarrow \infty} \langle x^*, x^* - x_n \rangle \leq 0.$$

Note that $\|y_n - x_n\| \rightarrow 0$. We derive that

$$(3.16) \quad \limsup_{n \rightarrow \infty} \langle x^*, x^* - y_n \rangle \leq 0.$$

Finally, we will show that $x_n \rightarrow x^*$. First, we have

$$(3.17) \quad \begin{aligned} \langle Tx_n - x^*, x_{n+1} - x^* \rangle &= \langle Tx_n - x^*, x_n - x^* \rangle + \langle Tx_n - x^*, x_{n+1} - x_n \rangle \\ &\leq \|x_n - x^*\|^2 + \|Tx_n - x^*\|\|x_{n+1} - x_n\|, \end{aligned}$$

and

$$\begin{aligned} \|y_n - x^*\|^2 &= \langle y_n - (1-\alpha_n)x_n, y_n - x^* \rangle + \langle (1-\alpha_n)x_n - x^*, y_n - x^* \rangle \\ &\leq \langle (1-\alpha_n)x_n - x^*, y_n - x^* \rangle \\ &= (1-\alpha_n)\langle x_n - x^*, y_n - x^* \rangle - \alpha_n\langle x^*, y_n - x^* \rangle \\ &\leq \frac{(1-\alpha_n)}{2}\|x_n - x^*\|^2 + \frac{1}{2}\|y_n - x^*\|^2 - \alpha_n\langle x^*, y_n - x^* \rangle. \end{aligned}$$

Thus,

$$(3.18) \quad \|y_n - x^*\|^2 \leq (1-\alpha_n)\|x_n - x^*\|^2 - 2\alpha_n\langle x^*, y_n - x^* \rangle.$$

Therefore, from (3.8), (3.13), (3.17) and (3.18), we get

$$\begin{aligned} \|x_{n+1} - x^*\|^2 &= \|(1-\beta_n)(y_n - x^*) + \beta_n(Tx_n - x^*)\|^2 \\ &\leq \|(1-\beta_n)(y_n - x^*)\|^2 + 2\beta_n\langle Tx_n - x^*, x_{n+1} - x^* \rangle \end{aligned}$$

$$\begin{aligned}
&\leq (1 - \beta_n)^2(1 - \alpha_n)\|x_n - x^*\|^2 - 2\alpha_n(1 - \beta_n)^2\langle x^*, y_n - x^* \rangle \\
&\quad + 2\beta_n\|x_n - x^*\|^2 + 2\beta_n\|Tx_n - x^*\|\|x_{n+1} - x_n\| \\
&\leq [1 - (1 - 2\beta_n)\alpha_n]\|x_n - x^*\|^2 + 2\alpha_n(1 - \beta_n)^2\langle x^*, x^* - y_n \rangle \\
&\quad + \beta_n^2\|x_n - x^*\|^2 + 2\beta_n\|Tx_n - x^*\|(L + 2)(\alpha_n + \beta_n)M_1 \\
(3.19) \quad &= (1 - \gamma_n)\|x_n - x^*\|^2 + \gamma_n\delta_n,
\end{aligned}$$

where $\gamma_n = (1 - 2\beta_n)\alpha_n$ and

$$\begin{aligned}
\delta_n &= \frac{2(1 - \beta_n)^2}{1 - 2\beta_n}\langle x^*, x^* - y_n \rangle + \frac{\beta_n^2}{(1 - 2\beta_n)\alpha_n}\|x_n - x^*\|^2 \\
&\quad + \frac{2\beta_n}{1 - 2\beta_n}\|Tx_n - x^*\|(L + 2)M_1 + \frac{2\beta_n^2}{(1 - 2\beta_n)\alpha_n}\|Tx_n - x^*\|(L + 2)M_1.
\end{aligned}$$

It is clear that $\sum_{n=0}^{\infty} \gamma_n = \infty$ and $\limsup_{n \rightarrow \infty} \delta_n \leq 0$. We can therefore apply Lemma 2.2 to (3.19) and conclude that $x_n \rightarrow x^*$ as $n \rightarrow \infty$. This completes the proof. \square

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Manuscript received September 18, 2013
revised March 2, 2014

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