

**A RELAXED PICARD ITERATION PROCESS
 FOR SET-VALUED OPERATORS OF THE MONOTONE TYPE**

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ABSTRACT. The fixed points \bar{x} of set-valued operators, $T: X \rightarrow 2^X$, satisfying a condition of the monotonicity type on convex subsets X of a Hilbert space are approximated by a relaxation process, $x_{n+1} = x_n + \omega_n(Tx_n - x_n)$, in which T is a single-valued branch of T and the relaxation parameter $\omega_n \in [0, 1]$ is made to depend in a certain way on the prior history of the process. If \tilde{T} is bounded on bounded subsets of X , then $\|x_n - \bar{x}\|$ converges to 0 like $O(n^{-1/2})$. If \tilde{T} is also continuous at \bar{x} and if $\bar{x} = \tilde{T}\bar{x}$, then $\|x_n - \bar{x}\| = o(n^{-1/2})$. If \tilde{T} satisfies a condition of the Lipschitz type at \bar{x} , then $\|x_n - \bar{x}\| = O(\mu^{n/2})$ for some $\mu \in [0, 1)$.

1. Introduction. Let X be a nonempty convex subset of a Hilbert space H , and suppose that for some $\bar{x} \in X$, the set-valued map $T: X \rightarrow 2^X$ satisfies the monotonicity condition,

$$1 > C_* \triangleq \sup_{\substack{x \in X \\ x \neq \bar{x}}} \sup_{\xi \in Tx} \frac{\operatorname{Re}\langle \xi - \bar{x}, x - \bar{x} \rangle}{\|x - \bar{x}\|^2}; \quad (1.1)$$

(every fixed point of T necessarily coincides with \bar{x}). Suppose also that the range of T is bounded in the sense that

$$\operatorname{diam} \bigcup_{x \in X} Tx < \infty \quad (1.2)$$

and therefore,

$$\sup_{x \in X} \sup_{\xi \in Tx} \|\xi - \bar{x}\| < \infty. \quad (1.3)$$

If \tilde{T} is any single-valued branch of T and if $\{x_n\} \subset X$ is generated by the associated relaxation process,

$$x_{n+1} = x_n + \omega_n(\tilde{T}x_n - x_n), \quad x_0 \in X, \quad (1.4)$$

where the relaxation parameter sequence $\{\omega_n\} \subset [0, 1]$ satisfies

$$\sum_{n=0}^{\infty} \omega_n^2 < \infty \quad (1.5A)$$

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and

$$\sum_{n=0}^{\infty} \omega_n = \infty. \quad (1.5B)$$

Then $\{x_n\}$ converges strongly to \bar{x} . Moreover, for certain special parameter sequences of this type, one can show that $\|x_n - \bar{x}\| = O(n^{-1/2})$ [1], [2], [3] (for related results see [4] also). To construct such a sequence, it is enough to have a value of C satisfying $1 > C \geq C_*$; no additional information about T or the course of the iteration (1.4) is required, hence the parameter sequence $\{\omega_n\}$ can be computed *before* (1.4) commences.

Roughly speaking, condition (1.5) requires that $\{\omega_n\}$ converge to 0 but not "too quickly". This restriction is necessary if (1.4) is to work in the worst cases where \tilde{T} is discontinuous at \bar{x} . Without continuity, it is possible for the residual norms $\|\tilde{T}x_n - x_n\|$ to remain bounded away from zero while x_n converges to a fixed point \bar{x} of \tilde{T} , and this behavior is generally incompatible with (1.4) unless $\omega_n \rightarrow 0$. Unfortunately, while (1.5) insures convergence in the worst cases, it virtually guarantees a uniformly poor asymptotic rate of convergence in *all* cases, even where \tilde{T} is Lipschitz continuous. On the other hand, it is known that for Lipschitz continuous \tilde{T} satisfying (1.1), the sequences $\{x_n\} \subset X$ generated by (1.4) will converge *geometrically*, or linearly, to \bar{x} (i.e., $\|x_n - \bar{x}\| = O(\lambda^n)$ for some $\lambda \in [0, 1)$), provided the terms in $\{\omega_n\}$ are sufficiently small but at the same time bounded away from 0 [5]–[9]. In fact, let \tilde{T} satisfy the condition

$$K_* \triangleq \sup_{\substack{x \in X \\ x \neq \bar{x}}} \frac{\|\tilde{T}x - \bar{x}\|}{\|x - \bar{x}\|} < \infty. \quad (1.6)$$

For $\omega_n \in [0, 1]$, (1.1) and (1.4) then give

$$\|x_{n+1} - \bar{x}\|^2 \leq \lambda_*(\omega_n) \|x_n - \bar{x}\|^2 \leq \lambda(\omega_n) \|x_n - \bar{x}\|^2 \quad (1.7A)$$

with

$$\lambda_*(\omega) = 1 - 2(1 - C_*)\omega + (1 - 2C_* + K_*^2)\omega^2 \quad (1.7B)$$

and

$$\lambda(\omega) = 1 - 2(1 - C)\omega + (1 - 2C + K^2)\omega^2 \quad (1.7C)$$

where

$$1 > C \geq C_* \quad (1.7D)$$

and

$$K \geq K_* \quad (1.7E)$$

If for all $n > 0$, $\omega_n = \omega$ with ω positive and sufficiently small, the parameter $\lambda(\omega_n)$ (and a fortiori $\lambda_*(\omega_n)$) will be nonnegative and smaller than 1, hence $\{x_n\}$ will converge to \bar{x} geometrically. Furthermore for any given bounds

$C \geq C_*$ and $K \geq K_*$, there is a "best" value $\omega = \omega_{OPT}$ which minimizes the parameter $\lambda(\omega)$ in the bound (1.7A). To see this, observe first that for fixed K and fixed $\omega \in [0, 1]$, $\lambda(\omega)$ increases with increasing C . Furthermore, since the Schwarz inequality, (1.1) and (1.6) give $|C_*| \leq K_*$, it follows that $K \geq K_*$ is always a better estimate of C_* than any $C > K$. Therefore, consideration may be limited to C 's in the range,

$$|C| \leq K \text{ and } C < 1. \tag{1.7F}$$

In this event, the coefficient of ω^2 in (1.7C) is always positive and $\lambda(\omega)$ attains its least value over $\omega \in [0, 1]$ at

$$\omega_{OPT} = \min \left\{ 1, \frac{1 - C}{1 - 2C + K^2} \right\} \tag{1.8A}$$

where

$$\lambda_{OPT} = \lambda(\omega_{OPT}) = \begin{cases} K^2, & \text{if } (1 - C) \geq 1 - 2C + K^2, \\ \frac{(K^2 - C^2)}{(1 - C)^2 + (K^2 - C^2)}, & \text{if } (1 - C) < 1 - 2C + K^2. \end{cases} \tag{1.8B}$$

In either case,

$$0 \leq \lambda_{*OPT} \leq \lambda_{OPT} < 1 \tag{1.8C}$$

where ω_{*OPT} and λ_{*OPT} are obtained by putting $C = C_*$ and $K = K_*$ in (1.8A) and (1.8B). These estimates show that asymptotic convergence rates far better than $O(n^{-1/2})$ are attainable with (1.4) in principle when \tilde{T} is sufficiently continuous at \bar{x} . However, in order to apply the relaxation parameter rule (1.8), one must have values for C and K which satisfy (1.7D) and (1.7E); if (1.8) is used with a $K < K_*$ or $C \notin [C_*, 1)$, then $\tilde{T}_{\omega_{OPT}} = I + \omega_{OPT}(\tilde{T} - I)$ is not necessarily contractive and the corresponding sequence $\{x_n\}$ generated by (1.4) with $\omega_n = \omega_{OPT}$ may not converge to any limit.

The present article considers a family of iterative processes (1.4) in which ω_n is made to depend in a simply way on the prior history of x_k , for $0 \leq k \leq n$. If T is set-valued and satisfies the monotonicity condition (1.1), the corresponding error norms $\|x_n - \bar{x}\|$ are of order $O(n^{-1/2})$ as $n \rightarrow \infty$ for any branch \tilde{T} that is bounded on bounded subsets of X . T need not satisfy the stronger boundedness condition (1.2) invoked in [1] and need not be continuous for this estimate to apply, however if \tilde{T} is continuous at \bar{x} and if $\bar{x} = \tilde{T}\bar{x}$, it turns out that $\|x_n - \bar{x}\| = o(n^{-1/2})$. Moreover, if \tilde{T} satisfies condition (1.6) at \bar{x} , then the error norms converge to 0 *geometrically*, even though neither the constant K_* nor an upper bound K for K_* actually appears in the rule which determines ω_n (and is therefore not required for an implementation of this rule).

In [10], Williamson considers a relaxation process (1.4) for $T: X \rightarrow 2^H$, with

ω_n depending on x_n according to a rule which turns out to be essentially a special case of the scheme proposed here; strong convergence of x_n to \bar{x} is established when X is dense in H (or in a solid sphere $\subset H$), and when T is bounded on bounded subsets of X . Useful a posteriori error estimates are also derived in [10], along with the basic $O(n^{-1/2})$ asymptotic convergence rate estimate.

The principal analytical tools employed in the next section were developed in [11] for a different purpose.

2. Results. Put

$$R_n = \tilde{T}x_n - x_n \quad \text{and} \quad E_n = x_n - \bar{x}.$$

Then (1.4) yields

$$E_{n+1} = E_n + \omega_n R_n$$

and it follows easily from the monotonicity condition (1.1) that for all $n > 0$ and $\omega_n > 0$,

$$\frac{\|E_{n+1}\|^2}{\theta^2} < [1 - 2(1 - C)\omega_n] \frac{\|E_n\|^2}{\theta^2} + \frac{\omega_n^2}{\theta^2} \|R_n\|^2 \quad (2.1)$$

where $\theta^2 \neq 0$ is a scaling parameter about which more will be said later, and where $1 > C > C_*$.

Let

$$\bar{\omega} = \min \left\{ 1, \frac{1}{2(1 - C)} \right\} > 0 \quad (2.2A)$$

and suppose that

$$0 < \omega_n \leq \bar{\omega} < 1 \quad (2.2B)$$

for all $n > 0$. Then the iterates $\{x_n\}$ generated by (1.4) cannot leave the convex set X . Moreover, the bracketed term on the right side of (2.1) is never negative and consequently a simple induction yields

$$0 < \|E_n\|^2 < B\beta_n \quad (2.3A)$$

for all $n > 0$, where,

$$B = \max \{ \theta^2, \|E_0\| \} \quad (2.3B)$$

and the sequence $\{\beta_n\} \subset [0, \infty)$ is recursively generated by

$$\beta_{n+1} = [1 - 2(1 - C)\omega_n] \beta_n + \frac{\omega_n^2}{\theta^2} \|R_n\|^2, \quad \beta_0 = 1. \quad (2.3C)$$

Since the error estimates (2.3) hold for *any* step length sequence $\{\omega_n\}$

satisfying (2.2), they are valid when ω_n is determined by the rule,

$$\omega_n = \tilde{\omega}(x_n, \beta_n; \theta) = \begin{cases} 0, & \text{if } R_n = 0, \\ \min\left\{\bar{\omega}, \frac{\theta^2(1-C)\beta_n}{\|R_n\|^2}\right\}, & \text{if } R_n \neq 0. \end{cases} \quad (2.4)$$

This scheme insures that $\{x_n\}$ will terminate at x_N if x_N is a fixed point of \tilde{T} , and otherwise minimizes the right side of (2.3C) over $\omega_n \in [0, \bar{\omega}]$, given β_n and R_n .

LEMMA 2.1. *If the sequences $\{\omega_n\}$, $\{\beta_n\}$, and $\{\|R_n\|\}$ satisfy (2.3C) and (2.4) for $n \geq 0$, then $\{\beta_n\}$ is nonnegative and monotone nonincreasing.*

PROOF. It has already been shown that $\beta_n \geq 0$ for all $n \geq 0$. When $\omega_n = 0$ in (2.4), (2.3C) $\Rightarrow \beta_{n+1} = \beta_n$; when $\omega_n = \theta^2(1-C)\beta_n/\|R_n\|^2$,

$$(2.3C) \Rightarrow \beta_{n+1} = \left[1 - \frac{\theta^2(1-C)^2\beta_n}{\|R_n\|^2}\right] \beta_n < \beta_n;$$

finally, when $\omega_n = \bar{\omega}$,

$$(2.3C) \Rightarrow \beta_{n+1} = [1 - 2(1-C)\bar{\omega}] \beta_n + \bar{\omega}^2 \frac{\|R_n\|^2}{\theta^2} < [1 - (1-C)\bar{\omega}] \beta_n < \max\left\{C, \frac{1}{2}\right\} \beta_n < \beta_n.$$

In all cases, $\beta_{n+1} < \beta_n$. Q.E.D.

COROLLARY 1. *Let $T: X \rightarrow 2^X$ satisfy the monotonicity condition (1.1), let the single-valued branch \tilde{T} in (1.4) be bounded on bounded subsets of X , and let $\{x_n\} \subset X$ be generated by (1.4)–(2.4). Then $\{x_n\}$ and the associated residual sequence $\{R_n\}$ are bounded.*

PROOF. $\{\beta_n\}$ is monotone nonincreasing and (2.3A) and (2.3B) hold, therefore $\|x_n - \bar{x}\| < B^{1/2}$ for all $n \geq 0$. Since T is bounded on bounded subsets, the sequence $\{\tilde{T}x_n - \bar{x}\}$ is also bounded. Boundedness of $\{Tx_n - x_n\}$ is now immediate from the triangle inequality. Q.E.D.

LEMMA 2.2. *Let $\{\beta_n\} \subset (0, \infty)$ and $\{q_n\} \subset [0, \infty)$ satisfy*

$$\beta_{n+1} \leq \beta_n - q_n\beta_n^2; \quad \beta_0 = 1 \quad (2.5)$$

for all $n \geq 0$. If $q_n \geq q > 0$ for $n \geq 0$, then

$$0 < \beta_n \leq \frac{1}{1 + qn} = O(n^{-1}). \quad (2.6)$$

Furthermore, if $\lim_{n \rightarrow \infty} q_n = \infty$, then $\beta_n = o(n^{-1})$.

PROOF.¹ Put $\delta_k = 1/\beta_k$. Then for all $k \geq 0$, one has $\delta_k - q_k > 0$ and $\delta_{k+1} - \delta_k \geq q_k \delta_k / (\delta_k - q_k) \geq q_k$. Consequently, for all $n \geq 0$,

$$\delta_n = \delta_0 + \sum_{k=0}^{n-1} (\delta_{k+1} - \delta_k) \geq 1 + \sum_{k=0}^{n-1} q_k.$$

If $q_k \geq q$ for all k , one then obtains (2.6). If $q_n \rightarrow \infty$, then for any $M > 0$ there is an L so large that $k \geq L \Rightarrow q_k \geq M$. For $n > L$ one then has

$$0 < n\beta_n = \frac{n}{\delta_n} \\ \leq \frac{n}{1 + \sum_{k=0}^{L-1} q_k + \sum_{k=L}^{n-1} q_k} \leq \frac{n}{(n-L)M}$$

and therefore $0 < \lim_{n \rightarrow \infty} n\beta_n \leq \overline{\lim}_n n\beta_n \leq 1/M$. Since M can be arbitrarily large, it follows that $\lim_{n \rightarrow \infty} n\beta_n = 0$. Q.E.D.

THEOREM 2.1. Let X be a nonempty convex subset of a Hilbert space H , let $T: X \rightarrow 2^X$ satisfy (1.1), and let \tilde{T} be a single-valued branch of T that is bounded on bounded subsets of X . Suppose that $\{x_n\} \subset X$, $\{\beta_n\} \subset [0, \infty)$, and $\{\omega_n\} \subset [0, \bar{\omega}]$ are generated by (1.4), (2.3C), and (2.4) with any $\theta \neq 0$, $C \in [C^*, 1)$, and $\bar{\omega} = \min\{1, 1/2(1-C)\}$. Let $\{R_n\}$ and $\{E_n\}$ denote the associated residual sequence and error sequence in (2.1). Then $\{R_n\}$ is bounded and for all $n \geq 0$,

$$0 \leq \|E_n\|^2 \leq \max\{\theta^2, \|E_0\|^2\} \cdot \beta_n \quad (2.7)$$

with $\{\beta_n\}$ nonnegative and monotone nonincreasing. Furthermore, either there is an $N > 0$ such that

$$x_n = \bar{x}, \quad \forall n \geq N \quad (2.8)$$

or else

$$0 < \beta_n \leq \frac{1}{1 + qn}, \quad \forall n \geq 0 \quad (2.9A)$$

with

$$q = \min\left\{A, \frac{\theta^2(1-C)^2}{d^2}\right\} \quad (2.9B)$$

$$A = 1 - \max\left\{\frac{1}{2}, C\right\} \quad (2.9C)$$

and

$$\infty > d \geq d_* \triangleq \sup_{n > 0} \|R_n\| \quad (2.9D)$$

in which case $\|E_n\| = O(n^{-1/2})$. Finally, if \bar{x} is a fixed point of \tilde{T} and if \tilde{T} is continuous at \bar{x} , then $\|x_n - \bar{x}\| = o(n^{-1/2})$.

¹This result is established in [7]; for convenience, the proof is repeated here.

PROOF. Corollary 1 of Lemma 2.1 $\Rightarrow \{R_n\}$ is bounded. (2.7) has already been established in (2.3). Lemma 2.1 $\Rightarrow \{\beta_n\}$ nonnegative and monotone nonincreasing. If $R_n = 0$, then x_n is a fixed point of \tilde{T} and must coincide with \bar{x} because of (1.1) [1]. Furthermore, (1.4)–(2.4) $\Rightarrow \omega_n = 0$ and $x_{n+1} = x_n = \bar{x}$. (2.8) now follows by induction.

If $R_n \neq 0$ and $\beta_n > 0$, then (2.3C) and (2.4) give

$$0 < \beta_{n+1} \leq (1 - q_n \beta_n) \beta_n \tag{2.10A}$$

with

$$q_n = \min \left\{ \frac{A}{\beta_n}, \frac{\theta^2 (1 - C)^2}{\|R_n\|^2} \right\} \geq q. \tag{2.10B}$$

Since $\beta = 1$, one obtains $\beta_n > 0$ for all $n \geq 0$ by induction if $R_n \neq 0$ for all $n \geq 0$. The estimate (2.9) is now immediate from Lemma 2.2.

Finally if $\bar{x} = \tilde{T}\bar{x}$, if \tilde{T} is continuous at \bar{x} , and if $R_n \neq 0$ for all $n \geq 0$, then (2.7) and (2.9) $\Rightarrow \|R_n\| \rightarrow 0$ and $\beta_n \rightarrow 0$ through positive values, in which case (2.10B) $\Rightarrow \lim_{n \rightarrow \infty} q_n = \infty$. It now follows from (2.7), (2.10A) and Lemma 2.2 that $\|E_n\| = o(n^{-1/2})$. Q.E.D.

THEOREM 2.2. *Let the hypothesis of Theorem 2.1 hold and suppose in addition that \tilde{T} satisfies (1.6). If $R_n \neq 0$ for $0 \leq n < N$, and $R_N = 0$, then (2.8) holds, and for $0 \leq n \leq N$,*

$$0 < \beta_n \leq \mu^n, \tag{2.11A}$$

with

$$\mu = 1 - \min \left\{ A, \frac{\theta^2 (1 - C)}{B(K_* + 1)^2} \right\} \leq 1 - \min \left\{ A, \frac{\theta^2 (1 - C)}{B(K + 1)^2} \right\} < 1, \tag{2.11B}$$

$$A = 1 - \max \left\{ \frac{1}{2}, C \right\}, \tag{2.11C}$$

$$B = \max \{ \theta^2, \|E_0\|^2 \} \tag{2.11D}$$

and

$$K \geq K_*. \tag{2.11E}$$

If $R_n \neq 0$ for all $n \geq 0$, then (2.11) holds for all $n \geq 0$ and it follows from (2.7) that $\|E_n\| = O(\mu^{n/2})$.

PROOF. If $R_n \neq 0$ for $0 \leq n < N$, then β_n is positive and satisfies (2.10) for $0 \leq n < N$. Furthermore,

$$\begin{aligned} (1.6) \Rightarrow \|R_n\|^2 &= \|Tx_n - x_n\|^2 \\ &\leq (\|Tx_n - \bar{x}\| + \|x_n - \bar{x}\|)^2 \leq (K + 1)^2 \|E_n\|^2 \end{aligned}$$

and therefore (2.7) and (2.10) give

$$\frac{\theta^2(1 - C)^2\beta_n}{\|R_n\|^2} \geq \frac{\theta^2(1 - C)^2}{B(K + 1)^2}$$

and

$$0 < 1 - q_n\beta_n \leq \mu$$

for $0 < n \leq N$. Q.E.D.

Note 2.1. Theorems 2.1 and 2.2 hold for any nonzero value of the scaling parameter θ . If E is any upper bound on $\|E_0\|$, then (2.7) and (2.8) give the estimate

$$\|E_n\|^2 \leq \kappa(\theta) \cdot n^{-1}$$

with

$$\kappa(\theta) = \max\{\theta^2, E^2\} \max\left\{\frac{1}{A}, \frac{d^2}{\theta^2(1 - C)^2}\right\}.$$

The coefficient $\kappa(\theta)$ always attains its least value when $\theta^2 = E^2$; for this value of θ , one has

$$\|E_n\|^2 \leq \max\left\{\frac{E^2}{A}, \frac{d^2}{(1 - C)^2}\right\} \cdot n^{-1}.$$

Similarly, if E bounds $\|E_0\|$, then (2.7) and (2.11) yield

$$\|E_n\|^2 \leq \max\{\theta^2, E^2\} (\bar{\mu}(\theta))^n$$

with

$$\bar{\mu}(\theta) = 1 - \min\left\{A, \frac{(1 - C)}{(K_* + 1)^2 \max\{E^2/\theta^2, 1\}}\right\}.$$

Once again, $\theta^2 = E^2$ yields the sharpest of these bounds, namely

$$\|E_n\|^2 \leq E^2 (\bar{\mu})^n$$

with

$$\bar{\mu} = 1 - \min\left\{A, (1 - C)/(K_* + 1)^2\right\}.$$

Note 2.2. If $X = H$ (or a linear variety in H) and if $T: X \rightarrow 2^X$, there is no reason for restricting ω_n to the interval $[0, 1]$, and Theorems 2.1 and 2.2 remain valid if $\bar{\omega}$ is replaced by $1/2(1 - C)$ in (2.4) and A is replaced by $\frac{1}{2}$ in (2.9).

Note 2.3. Williamson has shown in [10] that

$$\|E_0\| \leq E \stackrel{\Delta}{=} \|Ty - y\|/2(1 - C) \tag{2.12A}$$

if

$$x_0 = y + (Ty - y)/2(1 - C) \quad (2.12B)$$

for some $y \in X$. The estimate (2.12) can also be obtained by putting $\omega = 1/2(1 - C)$ in the inequality

$$\|x_0 - \bar{x}\|^2 \leq [1 - 2(1 - C)\omega] \|y - \bar{x}\|^2 + \omega^2 \|Ty - y\|^2, \quad \forall \omega > 0$$

which is gotten by the same argument that produces (2.1). For the case $X = H$, put $\bar{\omega} = 1/2(1 - C)$ in the rule (2.4) (see Note 2.2). Furthermore, let x_0 satisfy (2.12B) and set $\theta^2 = E^2$, with E given by (2.12A). Then (1.4)–(2.4) reduces to the basic iteration scheme in [10], with error bounds r_i related to the β_i 's in (2.3A) by

$$r_{i+1}^2 = E^2 \beta_i, \quad \forall i \geq 0$$

(see Note 2.1).

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