Review Article

Super-Relaxed ($\eta$)-Proximal Point Algorithms, Relaxed ($\eta$)-Proximal Point Algorithms, Linear Convergence Analysis, and Nonlinear Variational Inclusions

Ravi P. Agarwal$^{1,2}$ and Ram U. Verma$^{1,3}$

$^1$Department of Mathematical Sciences, Florida Institute of Technology, Melbourne, FL 32901, USA
$^2$Department of Mathematics and Statistics, King Fahd University of Petroleum and Minerals, Dhahran 31261, Saudi Arabia
$^3$International Publications (USA), 12085 Lake Cypress Circle, Suite 1109, Orlando, FL 32828, USA

Correspondence should be addressed to Ravi P. Agarwal, agarwal@fit.edu

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We glance at recent advances to the general theory of maximal (set-valued) monotone mappings and their role demonstrated to examine the convex programming and closely related field of nonlinear variational inequalities. We focus mostly on applications of the super-relaxed ($\eta$)-proximal point algorithm to the context of solving a class of nonlinear variational inclusion problems, based on the notion of maximal ($\eta$)-monotonicity. Investigations highlighted in this communication are greatly influenced by the celebrated work of Rockafellar (1976), while others have played a significant part as well in generalizing the proximal point algorithm considered by Rockafellar (1976) to the case of the relaxed proximal point algorithm by Eckstein and Bertsekas (1992). Even for the linear convergence analysis for the overrelaxed (or super-relaxed) ($\eta$)-proximal point algorithm, the fundamental model for Rockafellar’s case does the job. Furthermore, we attempt to explore possibilities of generalizing the Yosida regularization/approximation in light of maximal ($\eta$)-monotonicity, and then applying to first-order evolution equations/inclusions.

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1. Introduction and Preliminaries

We begin with a real Hilbert space $X$ with the norm $\| \cdot \|$ and the inner product $\langle \cdot, \cdot \rangle$. We consider the general variational inclusion problem of the following form. Find a solution to

$$0 \in M(x),$$

(1.1)

where $M : X \to 2^X$ is a set-valued mapping on $X$. 

In the first part, Rockafellar [1] introduced the proximal point algorithm, and examined the general convergence and rate of convergence analysis, while solving (1.1) by showing when \( M \) is maximal monotone, that the sequence \( \{x^k\} \) generated for an initial point \( x^0 \) by

\[
x^{k+1} \approx P_k(x^k)
\]

converges weakly to a solution of (1.1), provided that the approximation is made sufficiently accurate as the iteration proceeds, where \( P_k = (I + c_k M)^{-1} \) for a sequence \( \{c_k\} \) of positive real numbers that is bounded away from zero, and in second part using the first part and further amending the proximal point algorithm succeeded in achieving the linear convergence. It follows from (1.2) that \( x^{k+1} \) is an approximate solution to inclusion problem

\[
0 \in M(x) + c_k^{-1}(x - x^k).
\]

As a matter of fact, Rockafellar did demonstrate the weak convergence and strong convergence separately in two theorems, but for the strong convergence a further imposition of the Lipschitz continuity of \( M^{-1} \) at 0 plays the crucial part. Let us recall these results.

**Theorem 1.1** (see [1]). Let \( X \) be a real Hilbert space. Let \( M : X \to 2^X \) be maximal monotone, and let \( x^* \) be a zero of \( M \). Let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} \approx J_{c_k} M(x^k)
\]

such that

\[
\|x^{k+1} - J_{c_k} M(x^k)\| \leq \varepsilon_k,
\]

where \( J_{c_k} M = (I + c_k M)^{-1} \), \( \sum_{k=0}^{\infty} \varepsilon_k < \infty \), and \( \{c_k\} \) is bounded away from zero. Suppose that the sequence \( \{x^k\} \) is bounded in the sense that there exists at least one solution to \( 0 \in M(x) \).

Then the sequence \( \{x^k\} \) converges weakly to \( x^* \) for \( 0 \in M(x^*) \) with

\[
\lim_{k \to \infty} \|Q_k(x^k)\| = 0 \quad \text{for} \quad Q_k = I - J_{c_k} M.
\]

**Remark 1.2.** Note that Rockafellar [1] in Theorem 1.1, pointed out by a counterexample that the condition

\[
\sum_{k=0}^{\infty} \varepsilon_k < \infty
\]

is crucial; otherwise we may end up getting a nonconvergent sequence even with having just \( \varepsilon_k \to 0 \) and \( X \) one dimensional. Consider any maximal monotone mapping \( M \) such that the
set $T^{-1}(0) = \{x : 0 \in M(x)\}$, that is known always to be convex and contains more than one element. Then it turns out that $T^{-1}(x)$ contains a nonconvergent sequence $\{x^k\}$ such that

$$\|x^{k+1} - x^k\| \to 0,$$  \hspace{1cm} (1.8)

while

$$\sum_{k=0}^{\infty} \|x^{k+1} - x^k\| = \infty.$$  \hspace{1cm} (1.9)

The situation changes when $M = \partial f$ if the convex function $f$ attains its minimum non-uniquely.

Next we look, unlike Theorem 1.1, at [1, Theorem 2] in which Rockafellar achieved a linear convergence of the sequence by considering the Lipschitz continuity of $M^{-1}$ at 0 instead.

**Theorem 1.3** (see [1]). Let $X$ be a real Hilbert space. Let $M : X \to 2^X$ be maximal monotone, and let $x^*$ be a zero of $M$. Let the sequence $\{x^k\}$ be generated by the iterative procedure

$$x^{k+1} \approx J_{c_k}^M(x^k)$$  \hspace{1cm} (1.10)

such that

$$\|x^{k+1} - J_{c_k}^M(x^k)\| \leq \delta_k \|x^{k+1} - x^k\|,$$  \hspace{1cm} (1.11)

where $J_{c_k}^M = (I + c_k M)^{-1}$, $\sum_{k=0}^{\infty} \delta_k < \infty$, and $\{c_k\}$ is bounded away from zero. Suppose that the sequence $\{x^k\}$ is bounded in the sense that there exists at least one solution to $0 \in M(x)$. In addition, let $M^{-1}$ be Lipschitz continuous at 0 with modulus $a$, and

$$\mu_k = \frac{a}{(a^2 + c_k^2)^{1/2}} < 1.$$  \hspace{1cm} (1.12)

Then the sequence $\{x^k\}$ converges linearly to $x^*$ for $0 \in M(x^*)$ with

$$\|x^{k+1} - x^*\| \leq \theta_k \|x^k - x^*\| \quad \forall k \geq k',$$  \hspace{1cm} (1.13)

where

$$0 \leq \theta_k = \frac{\mu_k + \delta_k}{1 - \delta_k} < 1.$$  \hspace{1cm} (1.14)
Later on Rockafellar [1] applied Theorem 1.1 to a minimization problem regarding function $f : X \to (-\infty, +\infty]$, where $f$ is lower semicontinuous convex and proper by taking $M = \partial f$. It is well known that in this situation $\partial f$ is maximal monotone, and further

$$w \in \partial f(x) \iff f(x') \geq f(x) + \langle x' - x, w \rangle \quad \forall x'$$

or

$$\iff x \in \arg \min f - \langle \cdot, w \rangle.$$

As a specialization, we have

$$0 \in \partial f(x) \iff x \in \arg \min f.$$

That means, the proximal point algorithm for $M = \partial f$ is a minimizing method for $f$.

There is an abundance of literature on proximal point algorithms with applications mostly followed by the work of Rockafellar [1], but we focus greatly on the work of Eckstein and Bertsekas [2], where they have relaxed the proximal point algorithm in the following form and applied to the Douglas-Rachford splitting method. Now let us have a look at the relaxed proximal point algorithm introduced and studied in [2].

**Algorithm 1.4.** Let $M : X \to 2^X$ be a set-valued maximal monotone mapping on $X$ with $0 \in \text{range}(M)$, and let the sequence $\{x_k\}$ be generated by the iterative procedure

$$x^{k+1} = (1 - \alpha_k)x_k + \alpha_kw^k \quad \forall k \geq 0,$$

where $w^k$ is such that

$$\|w^k - (I + c_kM)^{-1}(x^k)\| \leq \epsilon_k \quad \forall k \geq 0,$$

$$\{\epsilon_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty)$$

are scalar sequences.

As a matter of fact, Eckstein and Bertsekas [2] applied Algorithm 1.4 to approximate a weak solution to (1.1). In other words, they established Theorem 1.1 using the relaxed proximal point algorithm instead.

**Theorem 1.5 (see [2, Theorem 3]).** Let $M : X \to 2^X$ be a set-valued maximal monotone mapping on $X$ with $0 \in \text{range}(M)$, and let the sequence $\{x^k\}$ be generated by Algorithm 1.4. If the scalar sequences $\{\epsilon_k\}, \{\alpha_k\}, \{c_k\}$ satisfy

$$E_1 = \sum_{k=0}^{\infty} \epsilon_k < \infty, \quad \Delta_1 = \inf \alpha_k > 0, \quad \Delta_2 = \sup \alpha_k < 2, \quad c = \inf c_k > 0,$$

then the sequence $\{x^k\}$ converges weakly to a zero of $M$. 

Convergence analysis for Algorithm 1.4 is achieved using the notion of the firm nonexpansiveness of the resolvent operator \((I + c_k M)^{-1}\). Somehow, they have not considered applying Algorithm 1.4 to Theorem 1.3 to the case of the linear convergence. The nonexpansiveness of the resolvent operator \((I + c_k M)^{-1}\) poses the prime difficulty to algorithmic convergence, and may be, this could have been the real steering for Rockafellar to the Lipschitz continuity of \(M^{-1}\) instead. That is why the Yosida approximation turned out to be more effective in this scenario, because the Yosida approximation

\[
M_{c_k} = c_k^{-1} \left[ I - (I + c_k M)^{-1} \right]
\]

(1.21)
takes care of the Lipschitz continuity issue.

As we look back into the literature, general maximal monotonicity has played a greater role to studying convex programming as well as variational inequalities/inclusions. Later it turned out that one of the most fundamental algorithms applied to solve these problems was the proximal point algorithm. In [2], Eckstein and Bertsekas have shown that much of the theory of the relaxed proximal point algorithm and related algorithms can be passed along to the Douglas-Rachford splitting method and its specializations, for instance, the alternating direction method of multipliers.

Just recently, Verma [3] generalized the relaxed proximal point algorithm and applied to the approximation solvability of variational inclusion problems of the form (1.1). Recently, a great deal of research on the solvability of inclusion problems is carried out using resolvent operator techniques, that have applications to other problems such as equilibria problems in economics, optimization and control theory, operations research, and mathematical programming.

In this survey, we first discuss in detail the history of proximal point algorithms with their applications to general nonlinear variational inclusion problems, and then we recall some significant developments, especially the relaxation of proximal point algorithms with applications to the Douglas-Rachford splitting method. At the second stage, we turn our attention to over-relaxed proximal point algorithms and their contribution to the linear convergence. We start with some introductory materials to the over-relaxed \((\eta)\)-proximal point algorithm based on the notion of maximal \((\eta)\)-monotonicity, and recall some investigations on approximation solvability of a general class of nonlinear inclusion problems involving maximal \((\eta)\)-monotone mappings in a Hilbert space setting. As a matter fact, we examine the convergence analysis of the over-relaxed \((\eta)\)-proximal point algorithm for solving a class of nonlinear inclusions. Also, several results on the generalized firm nonexpansiveness and generalized resolvent mapping are given. Furthermore, we explore the real impact of recently obtained results on the celebrated work of Rockafellar, most importantly in the case of over-relaxed (or super-relaxed) proximal point algorithms. For more details, we refer the reader [1–55].

We note that the solution set for (1.1) turns out to be the same as of the Yosida inclusion

\[
0 \in M_{\rho},
\]

(1.22)

where \(M_{\rho} = M(I + \rho M)^{-1}\) is the Yosida regularization of \(M\), while there is an equivalent form \(\rho^{-1}(I - (I + \rho M)^{-1})\), that is characterized as the Yosida approximation of \(M\) with
parameter $\rho > 0$. It seems in certain ways that it is easier to solve the Yosida inclusion than (1.1). In other words, $M_\rho$ provides better solvability conditions under right choice for $\rho$ than $M$ itself. To prove this assertion, let us recall the following existence theorem.

**Theorem 1.6.** Let $M : X \to 2^X$ be a set-valued maximal monotone mapping on $X$. Then the following statements are equivalent.

(i) An element $u \in X$ is a solution to $0 \in M_\rho(u)$.

(ii) $u = (I + \rho M)^{-1}(u)$.

Assume that $u$ is a solution to $0 \in M_\rho(u) = M(I + \rho M)^{-1}$. Then we have

$$0 \in M(I + \rho M)^{-1}(u)$$

$$\implies 0 \in \rho M(I + \rho M)^{-1}(u)$$

$$\implies (I + \rho M)^{-1}(u) \in (I + \rho M)(I + \rho M)^{-1}(u)$$

$$\implies u = (I + \rho M)^{-1}(u).$$

On the other hand, $M_\rho$ has also been applied to first-order evolution equations/inclusions in Hilbert space as well as in Banach space settings. As in our present situation, resolvent operator $(I + \rho M)^{-1}$ is empowered by $(\eta)$-maximal monotonicity, the Yosida approximation can be generalized in the context of solving first-order evolution equations/inclusions. In Zeidler [52, Lemma 31.7], it is shown that the Yosida approximation $M_\rho$ is $(2\rho^{-1})$-Lipschitz continuous, that is,

$$\|M_\rho(x) - M_\rho(y)\| \leq \frac{2}{\rho}\|x - y\| \quad \forall x, y \in D(M),$$

(1.24)

where this inequality is based on the nonexpansiveness of the resolvent operator $R^M_\rho = (I + \rho M)^{-1}$, though the result does not seem to be much application oriented, while if we apply the firm nonexpansiveness of the resolvent operator $R^M_\rho = (I + \rho M)^{-1}$, we can achieve, as applied in [5], more application-oriented results as follows:

$$\langle x - y, M_\rho(x) - M_\rho(y) \rangle \geq \rho\|M_\rho(x) - M_\rho(y)\|^2,$$

(1.25)

$$\|M_\rho(x) - M_\rho(y)\| \leq \frac{1}{\rho}\|x - y\| \quad \forall x, y \in D(M),$$

where the Lipschitz constant is $1/\rho$. 
Proof. For any \( x, y \in D(M) \), we have

\[
x - y = \rho(M_\rho(x) - M_\rho(y)) + R_\rho^M(x) - R_\rho^M(y).
\]  

(1.26)

Based on this equality and the firm nonexpansiveness of \( R_\rho^M \), we derive

\[
\langle x - y, M_\rho(x) - M_\rho(y) \rangle \\
= \left\langle \rho(M_\rho(x) - M_\rho(y)) + R_\rho^M(x) - R_\rho^M(y), M_\rho(x) - M_\rho(y) \right\rangle \\
= \rho \|M_\rho(x) - M_\rho(y)\|^2 + \left\langle R_\rho^M(x) - R_\rho^M(y), M_\rho(x) - M_\rho(y) \right\rangle \\
= \rho \|M_\rho(x) - M_\rho(y)\|^2 + \frac{1}{\rho} \left\langle R_\rho^M(x) - R_\rho^M(y), x - y - \left( R_\rho^M(x) - R_\rho^M(y) \right) \right\rangle \\
\geq \rho \|M_\rho(x) - M_\rho(y)\|^2 - \frac{1}{\rho} \|R_\rho^M(x) - R_\rho^M(y)\|^2 + \frac{1}{\rho} \|R_\rho^M(x) - R_\rho^M(y)\|^2 \\
= \rho \|M_\rho(x) - M_\rho(y)\|^2.
\]  

(1.27)

Thus, we have

\[
\langle x - y, M_\rho(x) - M_\rho(y) \rangle \geq \rho \|M_\rho(x) - M_\rho(y)\|^2. 
\]  

(1.28)

This completes the proof. \( \square \)

We note that from applications’ point of view, it seems that the result

\[
\langle x - y, M_\rho(x) - M_\rho(y) \rangle \geq \rho \|M_\rho(x) - M_\rho(y)\|^2,
\]  

(1.29)

that is, \( M_\rho \) is \( (\rho) \)-cocoercive, is relatively more useful than that of the nonexpansive form

\[
\|M_\rho(x) - M_\rho(y)\| \leq \frac{1}{\rho} \|x - y\| \quad \forall x, y \in D(M).
\]  

(1.30)

It is well known when \( M \) is maximal monotone, the resolvent operator \( R_\rho^M = (I + \rho M)^{-1} \) is single valued and Lipschitz continuous globally with the best constant \( \rho^{-1} \). Furthermore, the inverse resolvent identity is satisfied

\[
I - (I + \rho M)^{-1} = \left( I + (\rho M)^{-1} \right)^{-1}.
\]  

(1.31)

Indeed, the Yosida approximation \( M_\rho = \rho^{-1}(I - (I + \rho M)^{-1}) \) and its equivalent form \( M(I + \rho M)^{-1} \) are related to this identity. Let us consider

\[
\left( I - (I + \rho M)^{-1} \right) = \left( I + (\rho M)^{-1} \right)^{-1}.
\]  

(1.32)
Suppose that $u \in (I - (I + \rho M)^{-1})(w)$, then we have

$$u \in \left( I - (I + \rho M)^{-1} \right)(w) \quad \iff \quad u \in w - (I + \rho M)^{-1}(w) \quad \iff \quad w - u \in (I + \rho M)^{-1}(w) \quad \iff \quad w \in w - u + \rho M(w - u) \quad \iff \quad u \in \rho M(w - u) \quad \iff \quad w - u \in (\rho M)^{-1}(u) \quad \iff \quad w \in (I + (\rho M)^{-1})(u) \quad \iff \quad u \in \left( I + (\rho M)^{-1} \right)^{-1}(w).$$

(1.33)

On the other hand, we have the inverse resolvent identity that lays the foundation of the Yosida approximation.

**Lemma 1.7** (see [26, Lemma 12.14]). All mappings $M : X \to 2^X$ satisfy

$$\left( \rho I + M^{-1} \right)^{-1} = \rho^{-1} \left[ I - (I + \rho M)^{-1} \right] \text{ for } \rho > 0.$$  

(1.34)

**Proof.** We include the proof, though its similar to that of the above identity. Assume that $u \in \rho^{-1}(I - (I + \rho M)^{-1})(w)$, then we have

$$\rho u \in \left( I - (I + \rho M)^{-1} \right)(w) \quad \iff \quad \rho u \in w - (I + \rho M)^{-1}(w) \quad \iff \quad w - \rho u \in (I + \rho M)^{-1}(w) \quad \iff \quad w \in w - \rho u + \rho M(w - \rho u) \quad \iff \quad u \in M(w - \rho u) \quad \iff \quad w - \rho u \in M^{-1}(u) \quad \iff \quad w \in \left( \rho I + M^{-1} \right)(u) \quad \iff \quad u \in \left( \rho I + M^{-1} \right)^{-1}(w),$$

(1.35)

which is the required assertion. $\square$
Note that when $M : X \to 2^X$ is maximal monotone, mappings
\[
\left( I - (I + \rho M)^{-1} \right), \quad \left( I + (\rho M)^{-1} \right)^{-1}
\] are single valued, in fact maximal monotone and nonexpansive.

The contents for the paper are organized as follows. Section 1 deals with a general historical development of the relaxed proximal point algorithm and its variants in conjunction with maximal $(\eta)$-monotonicity, and with the approximation solvability of a class of nonlinear inclusion problems using the convergence analysis for the proximal point algorithm as well as for the relaxed proximal point algorithm. Section 2 introduces and derives some results on unifying maximal $(\eta)$-monotonicity and generalized firm nonexpansiveness of the generalized resolvent operator. In Section 3, the role of the over-relaxed $(\eta)$-proximal point algorithm is examined in detail in terms of its applications to approximating the solution of the inclusion problem (1.1). Finally, Section 4 deals with some important specializations that connect the results on general maximal monotonicity, especially to several aspects of the linear convergence.

2. General Maximal $\eta$-Monotonicity

In this section we discuss some results based on basic properties of maximal $\eta$-monotonicity, and then we derive some results involving $\eta$-monotonicity and the generalized firm nonexpansiveness. Let $X$ denote a real Hilbert space with the norm $\| \cdot \|$ and inner product $\langle \cdot, \cdot \rangle$. Let $M : X \to 2^X$ be a multivalued mapping on $X$. We will denote both the map $M$ and its graph by $M$, that is, the set\{(x,y) : y \in M(x)\}. This is equivalent to stating that a mapping is any subset $M$ of $X \times X$, and $M(x) = \{y : (x,y) \in M\}$. If $M$ is single valued, we will still use $M(x)$ to represent the unique $y$ such that $(x,y) \in M$ rather than the singleton set $\{y\}$. This interpretation will much depend on the context. The domain of a map $M$ is defined (as its projection onto the first argument) by

\[
\text{dom}(M) = \{x \in X : \exists y \in X : (x,y) \in M\} = \{x \in X : M(x) \neq \emptyset\}. \tag{2.1}
\]

$\text{dom}(T) = X$ will denote the full domain of $M$, and the range of $M$ is defined by

\[
\text{range}(M) = \{y \in X : \exists x \in X : (x,y) \in M\}. \tag{2.2}
\]

The inverse $M^{-1}$ of $M$ is $\{(y,x) : (x,y) \in M\}$. For a real number $\rho$ and a mapping $M$, let $\rho M = \{(x,\rho y) : (x,y) \in M\}$. If $L$ and $M$ are any mappings, we define

\[
L + M = \{(x,y+z) : (x,y) \in L, (x,z) \in M\}. \tag{2.3}
\]
Definition 2.1. Let $M : X \to 2^X$ be a multivalued mapping on $X$. The map $M$ is said to be

(i) monotone if

$$\langle u^* - v^*, u - v \rangle \geq 0 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(ii) $(r)$-strongly monotone if there exists a positive constant $r$ such that

$$\langle u^* - v^*, u - v \rangle \geq r\|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(iii) strongly monotone if

$$\langle u^* - v^*, u - v \rangle \geq \|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(iv) $(r)$-strongly pseudomonotone if

$$\langle v^*, u - v \rangle \geq 0$$

implies

$$\langle u^*, u - v \rangle \geq r\|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(v) pseudomonotone if

$$\langle v^*, u - v \rangle \geq 0$$

implies

$$\langle u^*, u - v \rangle \geq 0 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(vi) $(m)$-relaxed monotone if there exists a positive constant $m$ such that

$$\langle u^* - v^*, u - v \rangle \geq (-m)\|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(vii) cocoercive if

$$\langle u^* - v^*, u - v \rangle \geq \|u^* - v^*\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),$$

(viii) $(c)$-cocoercive if there is a positive constant $c$ such that

$$\langle u^* - v^*, u - v \rangle \geq c\|u^* - v^*\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M).$$
Definition 2.2. Let $M : X \to 2^X$ be a mapping on $X$. The map $M$ is said to be

(i) nonexpansive if

$$\|u^*-v^*\| \leq \|u-v\| \quad \forall (u,u^*), (v,v^*) \in \text{graph}(M),$$

(ii) firmly nonexpansive if

$$\|u^*-v^*\|^2 \leq \langle u^*-v^*, u-v \rangle \quad \forall (u,u^*), (v,v^*) \in \text{graph}(M),$$

(iii) $(c)$-firmly nonexpansive if there exists a constant $c > 0$ such that

$$\|u^*-v^*\|^2 \leq c\langle u^*-v^*, u-v \rangle \quad \forall (u,u^*), (v,v^*) \in \text{graph}(M).$$

In light of Definitions 2.1(vii) and 2.2(ii), notions of cocoerciveness and firm nonexpansiveness coincide, but differ in applications much depending on the context.

Definition 2.3. A map $\eta : X \times X \to X$ is said to be

(i) monotone if

$$\langle x-y, \eta(x,y) \rangle \geq 0 \quad \forall (x,y) \in X,$$

(ii) $(t)$-strongly monotone if there exists a positive constant $t$ such that

$$\langle x-y, \eta(x,y) \rangle \geq t\|x-y\|^2 \quad \forall (x,y) \in X,$$

(iii) strongly monotone if

$$\langle x-y, \eta(x,y) \rangle \geq \|x-y\|^2 \quad \forall (x,y) \in X,$$

(iv) $(\tau)$-Lipschitz continuous if there exists a positive constant $\tau$ such that

$$\|\eta(x,y)\| \leq \tau\|x-y\|.$$
Definition 2.4. Let \( M : X \to 2^X \) be a multivalued mapping on \( X \), and let \( \eta : X \times X \to X \) be another mapping. The map \( M \) is said to be

(i) \((\eta)\)-monotone if
\[
\langle u^* - v^*, \eta(u, v) \rangle \geq 0 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),
\] (2.21)

(ii) \((r, \eta)\)-strongly monotone if there exists a positive constant \( r \) such that
\[
\langle u^* - v^*, \eta(u, v) \rangle \geq r\|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),
\] (2.22)

(iii) \((\eta)\)-strongly monotone if
\[
\langle u^* - v^*, \eta(u, v) \rangle \geq \|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),
\] (2.23)

(iv) \((r, \eta)\)-strongly pseudomonotone if
\[
\langle v^*, \eta(u, v) \rangle \geq 0
\] implies
\[
\langle u^*, \eta(u, v) \rangle \geq r\|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),
\] (2.24)

(v) \((\eta)\)-pseudomonotone if
\[
\langle v^*, \eta(u, v) \rangle \geq 0
\] implies
\[
\langle u^*, \eta(u, v) \rangle \geq 0 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),
\] (2.25)

(vi) \((m, \eta)\)-relaxed monotone if there exists a positive constant \( m \) such that
\[
\langle u^* - v^*, \eta(u, v) \rangle \geq (-m)\|u - v\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M),
\] (2.26)

(vii) \((c, \eta)\)-cocoercive if there is a positive constant \( c \) such that
\[
\langle u^* - v^*, \eta(u, v) \rangle \geq c\|u^* - v^*\|^2 \quad \forall (u, u^*), (v, v^*) \in \text{graph}(M).
\] (2.27)
Definition 2.5. A map $M : X \to 2^X$ is said to be maximal $(\eta)$-monotone if

1. $M$ is $(\eta)$-monotone,
2. $R(I + cM) = X$ for $c > 0$.

Proposition 2.6. Let $\eta : X \times X \to X$ be a $(t)$-strongly monotone mapping, and let $M : X \to 2^X$ be a maximal $(\eta)$-monotone mapping. Then $(I + cM)$ is maximal $(\eta)$-monotone for $c > 0$, where $I$ is the identity mapping.

Proof. The proof follows on applying Definition 2.5.

Proposition 2.7 (see [4]). Let $\eta : X \times X \to X$ be $(t)$-strongly monotone, and let $M : X \to 2^X$ be maximal $(\eta)$-monotone. Then generalized resolvent operator $(I + cM)^{-1}$ is single valued, where $I$ is the identity mapping.

Proof. For a given $u \in X$, consider $x, y \in (I + cM)^{-1}(u)$ for $c > 0$. Since $M$ is maximal $(\eta)$-monotone, we have

$$
\frac{1}{c}[-x + u] \in M(x),
$$

(2.30)

$$
\frac{1}{c}[-y + u] \in M(y).
$$

Now using the $(\eta)$-monotonicity of $M$, it follows that

$$
\langle -x + u - (-y + u), \eta(x, y) \rangle = \langle y - x, \eta(x, y) \rangle \geq 0.
$$

(2.31)

Since $\eta$ is $(t)$-strongly monotone, it implies $x = y$. Thus, $(I + cM)^{-1}$ is single valued.

Definition 2.8. Let $\eta : X \times X \to X$ be $(t)$-strongly monotone, and let $M : X \to 2^X$ be maximal $(\eta)$-monotone. Then the generalized resolvent operator $J_{c, \eta}^M : X \to X$ is defined by

$$
J_{c, \eta}^M(u) = (I + cM)^{-1}(u) \quad \text{for } c > 0.
$$

(2.32)

Proposition 2.9 (see [4]). Let $X$ be a real Hilbert space, let $M : X \to 2^X$ be maximal $(\eta)$-monotone, and let $\eta : X \times X \to X$ be $(t)$-strongly monotone. Then the resolvent operator associated with $M$ and defined by

$$
J_{\rho, \eta}^M(u) = (I + \rho M)^{-1}(u) \quad \forall u \in X
$$

(2.33)

satisfies the following:

$$
\langle u - v, \eta(J_{\rho, \eta}^M(u), J_{\rho, \eta}^M(v)) \rangle \geq t \left\| J_{\rho, \eta}^M(u) - J_{\rho, \eta}^M(v) \right\|^2.
$$

(2.34)
Proof. For any \( u, v \in X \), it follows from the definition of the resolvent operator \( J_{\rho}^{M,\eta} \) that

\[
\frac{1}{\rho} \left[ u - J_{\rho}^{M,\eta}(u) \right] \in M \left( J_{\rho}^{M,\eta}(u) \right),
\]

\[
\frac{1}{\rho} \left[ v - J_{\rho}^{M,\eta}(v) \right] \in M \left( J_{\rho}^{M,\eta}(v) \right).
\]

Since \( M \) is \((\eta)\)-monotone, we have

\[
\frac{1}{\rho} \left[ u - v - \left[ J_{\rho}^{M,\eta}(u) - J_{\rho}^{M,\eta}(v) \right] \right] \in \eta \left( J_{\rho}^{M,\eta}(u), J_{\rho}^{M,\eta}(v) \right) \geq 0.
\]

In light of (2.36), we have

\[
\left\langle u - v, \eta \left( J_{\rho}^{M,\eta}(u), J_{\rho}^{M,\eta}(v) \right) \right\rangle \geq \left( J_{\rho}^{M,\eta}(u) - J_{\rho}^{M,\eta}(v) \right) \eta \left( J_{\rho}^{M,\eta}(u), J_{\rho}^{M,\eta}(v) \right) \geq t \left\| J_{\rho}^{M,\eta}(u) - J_{\rho}^{M,\eta}(v) \right\|^2
\]

(2.37)

\[\square\]

**Proposition 2.10** (see [4]). Let \( X \) be a real Hilbert space, let \( M : X \to 2^X \) be maximal \((\eta)\)-monotone, and let \( \eta : X \times X \to X \) be \((1)\)-strongly monotone.

If, in addition, (for \( \gamma > 0 \))

\[
\left\langle u - v, J_{\rho}^{M,\eta}(u) - J_{\rho}^{M,\eta}(v) \right\rangle \geq \gamma \left\langle u - v, \eta \left( J_{\rho}^{M,\eta}(u), J_{\rho}^{M,\eta}(v) \right) \right\rangle \forall u, v \in X,
\]

(2.38)

then, for \( J_k^* = I - J_{\rho}^{M,\eta} \), one has (for \( t \geq 1 \))

\[
\left\langle u - v, J_k^*(u) - J_k^*(v) \right\rangle \geq \frac{\gamma t - 1}{2\gamma t - 1} \left\| u - v \right\|^2 + \frac{\gamma t}{2\gamma t - 1} \left\| J_k^*(u) - J_k^*(v) \right\|^2,
\]

(2.39)

where

\[
J_{\rho}^{M,\eta}(u) = (I + \rho M)^{-1}(u) \forall u \in X.
\]

(2.40)
Proof. We include the proof for the sake of the completeness. To prove (2.39), we apply (2.38) to Proposition 2.9, and we get

\[ \langle u - v, J_p^{M,\eta}(u) - J_p^{M,\eta}(v) \rangle \geq \gamma t \| J_p^{M,\eta}(u) - J_p^{M,\eta}(v) \|^2. \] (2.41)

It further follows that

\[ \langle u - v, u - v - (J_k^*(u) - J_k^*(v)) \rangle \]

\[ \geq \gamma t \left[ \| J_k^*(u) - J_k^*(v) \|^2 + \| u - v \|^2 - 2 \langle J_k^*(u) - J_k^*(v), u - v \rangle \right]. \] (2.42)

□

When \( \gamma = 1 \) and \( t > 1 \) in Proposition 2.10, we have the following.

**Proposition 2.11.** Let \( X \) be a real Hilbert space, let \( M : X \rightarrow 2^X \) be maximal \((\eta)\)-monotone, and let \( \eta : X \times X \rightarrow X \) be \((t)\)-strongly monotone.

If, in addition, one supposes that

\[ \langle u - v, J_p^{M,\eta}(u) - J_p^{M,\eta}(v) \rangle \geq \langle u - v, \eta (J_p^{M,\eta}(u), J_p^{M,\eta}(v)) \rangle \quad \forall u, v \in X, \] (2.43)

then, for \( J_k^* = I - J_p^{M,\eta} \), one has (for \( t > 1 \))

\[ \langle u - v, J_k^*(u) - J_k^*(v) \rangle \geq \frac{t - 1}{2t - 1} \| u - v \|^2 + \frac{t}{2t - 1} \| J_k^*(u) - J_k^*(v) \|^2, \] (2.44)

where

\[ J_p^{M,\eta}(u) = (I + \rho M)^{-1}(u) \quad \forall u \in X. \] (2.45)

For \( t = 1 \) and \( \gamma > 1 \) in Proposition 2.10, we find a result of interest as follows.

**Proposition 2.12.** Let \( X \) be a real Hilbert space, let \( M : X \rightarrow 2^X \) be maximal \((\eta)\)-monotone, and let \( \eta : X \times X \rightarrow X \) be strongly monotone.

If, in addition, one supposes (for \( \gamma > 1 \)) that

\[ \langle u - v, J_p^{M,\eta}(u) - J_p^{M,\eta}(v) \rangle \geq \gamma \langle u - v, \eta (J_p^{M,\eta}(u), J_p^{M,\eta}(v)) \rangle \quad \forall u, v \in X, \] (2.46)

then, for \( J_k^* = I - J_p^{M,\eta} \), one has

\[ \langle u - v, J_k^*(u) - J_k^*(v) \rangle \geq \frac{\gamma - 1}{2\gamma - 1} \| u - v \|^2 + \frac{\gamma}{2\gamma - 1} \| J_k^*(u) - J_k^*(v) \|^2, \] (2.47)

where

\[ J_p^{M,\eta}(u) = (I + \rho M)^{-1}(u) \quad \forall u \in X. \] (2.48)

For \( \gamma = t = 1 \) in Proposition 2.10, we have the following result.
Proposition 2.13. Let $X$ be a real Hilbert space, let $M : X \to 2^X$ be maximal $(\eta)$-monotone, and let $\eta : X \times X \to X$ be strongly monotone.

If, in addition, one assumes that
\[
\langle u - v, J^{M,\eta}_\rho(u) - J^{M,\eta}_\rho(v) \rangle \geq \langle u - v, \eta(J^{M,\eta}_\rho(u), J^{M,\eta}_\rho(v)) \rangle \quad \forall u, v \in X, \tag{2.49}
\]
then, for $J^*_k = I - J^{M,\eta}_\rho$, one has
\[
\langle u - v, J^*_k(u) - J^*_k(v) \rangle \geq \| J^*_k(u) - J^*_k(v) \|^2, \tag{2.50}
\]
where
\[
J^{M,\eta}_\rho(u) = (I + \rho M)^{-1}(u) \quad \forall u \in X. \tag{2.51}
\]

3. The Over-Relaxed $(\eta)$-Proximal Point Algorithm

This section deals with the over-relaxed $(\eta)$-proximal point algorithm and its application to approximation solvability of the inclusion problem (1.1) based on the maximal $(\eta)$-monotonicity. Furthermore, some results connecting the $(\eta)$-monotonicity and corresponding resolvent operator are established, that generalize the results on the firm nonexpansiveness [2], while the auxiliary results on maximal $(\eta)$-monotonicity and general maximal monotonicity are obtained.

Theorem 3.1. Let $X$ be a real Hilbert space, and let $M : X \to 2^X$ be maximal $(\eta)$-monotone. Then the following statements are mutually equivalent.

(i) An element $u \in X$ is a solution to (1.1).

(ii) For an $u \in X$, one has
\[
u = J^{M,\eta}_c(u) \quad \text{for } c > 0, \tag{3.1}
\]
where
\[
J^{M,\eta}_c(u) = (I + c M)^{-1}(u). \tag{3.2}
\]

Proof. It follows from the definition of the generalized resolvent operator corresponding to $M$.

Note that Theorem 3.1 generalizes [2, Lemma 2] to the case of a maximal $(\eta)$-monotone mapping.

Next, we present a generalization to the relaxed proximal point algorithm [3] based on the maximal $(\eta)$-monotonicity.
Algorithm 3.2 (see [4]). Let $M : X \to 2^X$ be a set-valued maximal $(\eta)$-monotone mapping on $X$ with $0 \in \text{range}(M)$, and let the sequence $\{x^k\}$ be generated by the iterative procedure

$$x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,$$

(3.3)

and $y^k$ satisfies

$$\left\|y^k - J_{c_k}^{M,\eta}(x^k)\right\| \leq \delta_k \left\|y^k - x^k\right\|,$$

(3.4)

where $J_{c_k}^{M,\eta} = (I + c_k M)^{-1}$, $\delta_k \to 0$ and

$$y^{k+1} = (1 - \alpha_k)x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) \quad \forall k \geq 0.$$

(3.5)

Here

$$\{\delta_k\}, \{|\alpha_k|\}, \{|c_k|\} \subseteq [0, \infty)$$

(3.6)

are scalar sequences such that $\sum_{k=0}^{\infty} \delta_k < \infty$.

Algorithm 3.3. Let $M : X \to 2^X$ be a set-valued maximal $(\eta)$-monotone mapping on $X$ with $0 \in \text{range}(M)$, and let the sequence $\{x^k\}$ be generated by the iterative procedure

$$x^{k+1} = (1 - \alpha_k - \beta_k)x^k + \alpha_k y^k \quad \forall k \geq 0,$$

(3.7)

and $y^k$ satisfies

$$\left\|y^k - J_{c_k}^{M}(x^k)\right\| \leq \epsilon_k,$$

(3.8)

where $J_{c_k}^{M,\eta} = (I + c_k M)^{-1}$, and

$$\{\epsilon_k\}, \{|\alpha_k|\}, \{|\beta_k|\}, \{|c_k|\} \subseteq [0, \infty)$$

(3.9)

are scalar sequences such that $\sum_{k=0}^{\infty} \epsilon_k < \infty$.

For $\delta_k = 1/k^2$ in Algorithm 3.2, we have the following.
**Algorithm 3.4.** Let \( M : X \to 2^X \) be a set-valued maximal \((\eta)\)-monotone mapping on \( X \) with \( 0 \in \text{range}(M) \), and let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,
\]

and \( y^k \) satisfies

\[
\left\| y^k - J_{c_k}^{M,\eta}(x^k) \right\| \leq \frac{1}{k^2} \left\| y^k - x^k \right\|,
\]

where \( J_{c_k}^{M,\eta} = (I + c_k M)^{-1} \), and

\[
y^{k+1} = (1 - \alpha_k)x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) \quad \forall k \geq 0.
\]

Here

\[
\{\alpha_k\}, \{c_k\} \subseteq [0, \infty)
\]

are scalar sequences.

In the following result [4], we observe that Theorems 1.1 and 1.3 are unified and are generalized to the case of the \( \eta \)-maximal monotonicity and super-relaxed proximal point algorithm. Also, we notice that this result in certain respects demonstrates the importance of the firm nonexpansiveness rather than of the nonexpansiveness.

**Theorem 3.5** (see [4]). Let \( X \) be a real Hilbert space. Let \( M : X \to 2^X \) be maximal \((\eta)\)-monotone, and let \( x^* \) be a zero of \( M \). Let \( \eta : X \times X \to X \) be \((t)\)-strongly monotone. Furthermore, assume (for \( \gamma > 0 \))

\[
\langle u - v, J_{c_k}^{M,\eta}(u) - J_{c_k}^{M,\eta}(v) \rangle \geq \frac{1}{\gamma} \langle u - v, \eta(J_{c_k}^{M,\eta}(u), J_{c_k}^{M,\eta}(v)) \rangle \quad \forall u, v \in X.
\]

Let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,
\]

and \( y^k \) satisfies

\[
\left\| y^k - J_{c_k}^{M,\eta}(x^k) \right\| \leq \epsilon_k,
\]

where \( J_{c_k}^{M,\eta} = (I + c_k M)^{-1} \), \( \sum_{k=0}^{\infty} \epsilon_k < \infty \), \( \{\alpha_k\}, \{\epsilon_k\}, \{c_k\} \subseteq [0, \infty) \), \( c_k \not< \epsilon^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2\gamma t/(2\gamma t - 1) \).

Suppose that the sequence \( \{x^k\} \) is bounded in the sense that there exists at least one solution to \( 0 \in M(x) \).
Then one has (for \( t \geq 1 \))

\[
(2\gamma t - 1) \left\| J_{c_k}^{M,\eta}(x^k) - x^* \right\|^2 \leq \left\| x^k - x^* \right\|^2 - \left\| J_k^*(x^k) \right\|^2 ,
\]

(3.17)

where \( \gamma t > 1 \) and

\[
J_k^* = I - J_{c_k}^{M,\eta}.
\]

(3.18)

In addition, suppose that the sequence \( \{x^k\} \) is generated by Algorithm 3.2 as well, and that \( M^{-1} \) is \((a)\)-Lipschitz continuous at 0, that is, there exists a unique solution \( z^* \) to \( 0 \in M(z) \) (equivalently, \( M^{-1}(0) = \{z^*\} \)) and for constants \( a \geq 0 \) and \( b > 0 \), one has

\[
\|z - z^*\| \leq a\|w\| \quad \text{whenever} \quad z \in M^{-1}(w), \|w\| \leq b.
\]

(3.19)

Here

\[
\{\delta_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty)
\]

(3.20)

are scalar sequences such that \( \delta_k \to 0 \) and \( \sum_{k=0}^{\infty} \delta_k < \infty \).

Then the sequence \( \{x^k\} \) converges linearly to a unique solution \( x^* \) with rate

\[
\sqrt{1 - \alpha^* [2(1 - \gamma td^2) - \alpha^* (1 - (2\gamma t - 1)d^2)]} < 1 \quad \text{for} \quad t \geq 1,
\]

(3.21)

where \( d = \sqrt{a^2/(c^* + (2\gamma t - 1)a^2)} \), \( \alpha^* = \limsup_{k \to \infty} \alpha_k \), and sequences \( \{\alpha_k\} \) and \( \{c_k\} \) satisfy \( \alpha_k \geq 1 \), \( c_k \geq c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2\gamma t/(2\gamma t - 1) \).

Proof. Suppose that \( x^* \) is a zero of \( M \). For all \( k \geq 0 \), we set

\[
J_k^* = I - J_{c_k}^{M,\eta}.
\]

(3.22)

Therefore, \( J_k^*(x^*) = 0 \). Then, in light of Theorem 3.1, any solution to (1.1) is a fixed point of \( J_{c_k}^{M,\eta} \), and hence a zero of \( J_k^* \).

Next, the proof of (3.17) follows from a regular manipulation, and the following equality:

\[
\|u - v\|^2 = \left\| J_{c_k}^{M,\eta}(u) - J_{c_k}^{M,\eta}(v) + J_k^*(u) - J_k^*(v) \right\|^2 \quad \forall u, v \in X.
\]

(3.23)
Before we start establishing linear convergence of the sequence \( \{x^k\} \), we express \( \{x^k\} \) in light of Algorithm 3.2 as

\[
y^{k+1} = (1 - \alpha_k)x^k + \alpha_k J_{cs}^{M,\eta}(x^k)
= (I - \alpha_k J_0^\star)(x^k).
\] (3.24)

Now we begin verifying the boundedness of the sequence \( \{x^k\} \) leading to \( x^k - J_{cs}^{M,\eta}(x^k) \to 0 \).

Next, we estimate using Proposition 2.10 (for \( t \geq 1 \))

\[
\|y^{k+1} - x^*\|^2 = \|(1 - \alpha_k)x^k + \alpha_k J_{cs}^{M,\eta}(x^k) - x^*\|^2
= \|x^k - x^* - \alpha_k J_0^\star(x^k)\|^2
= \|x^k - x^*\|^2 - 2\alpha_k \langle x^k - x^*, J_0^\star(x^k) - J_0^\star(x^*) \rangle + \alpha_k^2 \|J_0^\star(x^k)\|^2
\leq \|x^k - x^*\|^2 - \frac{2(\gamma t - 1)\alpha_k}{2\gamma t - 1}\|x^k - x^*\|^2 - \frac{2\gamma t}{2\gamma t - 1}\alpha_k \|J_0^\star(x^k)\|^2 + \alpha_k^2 \|J_0^\star(x^k)\|^2
= \left(1 - \frac{2(\gamma t - 1)\alpha_k}{2\gamma t - 1}\right)\|x^k - x^*\|^2 - \alpha_k \left(\frac{2\gamma t}{2\gamma t - 1} - \alpha_k\right) \|J_0^\star(x^k)\|^2.
\] (3.25)

Since under the assumptions \( \alpha_k(2\gamma t/(2\gamma t - 1) - \alpha_k) > 0 \), it follows that

\[
\|y^{k+1} - x^*\| \leq \Delta \|x^k - x^*\| \leq \|x^k - x^*\|,
\] (3.26)

where \( \Delta = \sqrt{1 - 2(\gamma t - 1)\alpha_k/(2\gamma t - 1)} < 1 \).

Moreover,

\[
\|x^{k+1} - y^{k+1}\| = \|(1 - \alpha_k)x^k + \alpha_k y^k - [(1 - \alpha_k)x^k + \alpha_k J_{cs}^{M,\eta}(x^k)]\|
= \|\alpha_k (y^k - J_{cs}^{M,\eta}(x^k))\|
\leq \alpha_k \epsilon_k.
\] (3.27)
Now we find the estimate leading to the boundedness of the sequence \( \{x^k\} \).

\[
\|x^{k+1} - x^*\| \leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\| \\
\leq \|x^k - x^*\| + \alpha_k e_k \\
\leq \|x^0 - x^*\| + \sum_{j=0}^{k} \alpha_j e_j \tag{3.28}
\]

Thus, the sequence \( \{x^k\} \) is bounded.

We further examine the estimate

\[
\|x^{k+1} - x^*\|^2 = \|y^{k+1} - x^* + x^{k+1} - y^{k+1}\|^2 \\
= \|y^{k+1} - x^*\|^2 + 2 \langle y^{k+1} - x^*, x^{k+1} - y^{k+1} \rangle + \|x^{k+1} - y^{k+1}\|^2 \\
\leq \|y^{k+1} - x^*\|^2 + 2 \|y^{k+1} - x^*\| \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2 \\
\leq \|x^k - x^*\|^2 - \alpha_k \left( \frac{2\gamma t}{2\gamma t - 1} - \alpha_k \right) \|J^*_k(x^k)\|^2 + 2 \langle x^{k+1} - x^*, \|x^{k+1} - y^{k+1}\| \rangle \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2 \\
\leq \|x^k - x^*\|^2 - \alpha_k \left( \frac{2\gamma t}{2\gamma t - 1} - \alpha_k \right) \|J^*_k(x^k)\|^2 + 2 \left( \|x^0 - x^*\| + 2 \|x^{k+1} - y^{k+1}\| \right) \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2 \\
\leq \|x^0 - x^*\|^2 + 2 \|x^{k+1} - x^*\| \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\| \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2 \\
\leq \|x^0 - x^*\|^2 + \left( \frac{2\gamma t}{2\gamma t - 1} - \alpha_k \right) \|J^*_k(x^k)\|^2 + 2 \left( \|x^0 - x^*\| + \frac{2\gamma t}{2\gamma t - 1} \sum_{k=0}^{\infty} e_k \right) \|x^{k+1} - y^{k+1}\| + \frac{2\gamma t}{2\gamma t - 1} \sum_{k=0}^{\infty} e_k \|x^{k+1} - y^{k+1}\|^2,
\]

where \( \alpha_k (2\gamma t / (2\gamma t - 1) - \alpha_k) > 0 \).

Since \( \{e_k\} \) is summable, so is \( \{e_k^2\} \), and hence \( \sum_{k=0}^{\infty} e_k^2 < \infty \). As \( k \to \infty \), we have that

\[
\sum_{j=0}^{k} \|J^*_j(x^j)\|^2 < \infty \implies \lim_{k \to \infty} J^*_k(x^k) = 0, \tag{3.30}
\]

that is, \( x^k - J^*_k(x^k) \to 0 \).
Now we turn our attention (using the previous argument) to linear convergence of the sequence \( \{x^k\} \). Since \( \lim_{k \to \infty} J_k^*(x^k) = 0 \), it implies for \( k \) large that \( c_k^{-1} J_k^*(x^k) \in M(J_{c_k}^{M,\eta}(x^k)) \). Moreover, \( \|c_k^{-1} J_k^*(x^k)\| \leq b \) for \( k \geq k' \) and \( b > 0 \). Therefore, in light of (3.19), by taking \( w = c_k^{-1} J_k^*(x^k) \) and \( z = J_{c_k}^{M,\eta}(x^k) \), we have

\[
\|J_{c_k}^{M,\eta}(x^k) - x^*\| \leq a \|c_k^{-1} J_k^*(x^k)\| \quad \forall k \geq k'.
\]  

(3.31)

Applying (3.17), we arrive at

\[
\|J_{c_k}^{M,\eta}(x^k) - x^*\|^2 \leq \frac{a^2}{c_k^2 + (2\gamma t - 1)a^2} \|x^k - x^*\|^2 \quad \text{for} \ t \geq 1,
\]  

(3.32)

where \( J_{c_k}^{M,\eta}(x^*) = x^* \).

Since \( y^{k+1} : = (1 - \alpha_k)x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) \), we estimate using (3.32) and \( (\alpha_k \geq 1) \) that

\[
\|y^{k+1} - x^*\|^2
\]

\[
= \| (1 - \alpha_k)x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) - x^* \|^2
\]

\[
= \| \alpha_k (J_{c_k}^{M,\eta}(x^k) - x^*) + (1 - \alpha_k) (x^k - x^*) \|^2
\]

\[
= \alpha_k^2 \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 + 2\alpha_k (1 - \alpha_k) \langle J_{c_k}^{M,\eta}(x^k) - x^*, x^k - x^* \rangle
\]

\[
\leq \alpha_k^2 \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 + 2\gamma t \alpha_k (1 - \alpha_k) \| J_{c_k}^{M,\eta}(x^k) - x^* \| \| x^k - x^* \|
\]

\[
\leq \alpha_k^2 \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 + 2\gamma t \alpha_k (1 - \alpha_k) \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2
\]

\[
= [2\alpha_k (1 - \alpha_k) \gamma t + \alpha_k^2] \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2
\]

\[
= \alpha_k [2\gamma t - (2\gamma t - 1) \alpha_k] \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2
\]

\[
\leq \alpha_k [2\gamma t - (2\gamma t - 1) \alpha_k] \frac{a^2}{c_k^2 + (2\gamma t - 1)a^2} \| x^k - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2
\]

\[
= \left( \alpha_k [2\gamma t - (2\gamma t - 1) \alpha_k] \frac{a^2}{c_k^2 + (2\gamma t - 1)a^2} + (1 - \alpha_k)^2 \right) \| x^k - x^* \|^2,
\]  

(3.33)

where \( \alpha_k [2\gamma t - (2\gamma t - 1) \alpha_k] > 0 \).
Hence, we have
\[ \|y^{k+1} - x^*\| \leq \theta_k \|x^k - x^*\|, \quad (3.34) \]

where
\[ \theta_k = \sqrt{\alpha_k [2\gamma t - (2\gamma t - 1)\alpha_k] \left[ \frac{a^2}{c_k^2 + (2\gamma t - 1)a^2} \right] + (1 - \alpha_k)^2 < 1, \quad (3.35) \]

for \( \alpha_k [2\gamma t - (2\gamma t - 1)\alpha_k] > 0 \) and \( \alpha_k \geq 1. \)

Since Algorithm 3.2 ensures
\[ \|y^k - J_{c_k}^{M,\eta}(x^k)\| \leq \delta_k \|y^k - x^k\|, \]
\[ \alpha_k (y^k - x^k) = x^{k+1} - x^k, \quad (3.36) \]
we have
\[ \|x^{k+1} - y^{k+1}\| = \|\alpha_k \left( y^k - J_{c_k}^{M,\eta}(x^k) \right)\| \leq \alpha_k \delta_k \|y^k - x^k\|, \]
\[ \|x^{k+1} - x^*\| = \|y^{k+1} - x^* + x^{k+1} - y^{k+1}\| \leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\| \leq \|y^{k+1} - x^*\| + \alpha_k \delta_k \|y^k - x^k\| \]
\[ = \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^k\| \leq \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\| \leq \theta_k \|x^k - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\|. \quad (3.37) \]

It follows that
\[ \|x^{k+1} - x^*\| \leq \frac{\theta_k + \delta_k}{1 - \delta_k} \|x^k - x^*\|, \quad (3.38) \]

where
\[ \limsup \frac{\theta_k + \delta_k}{1 - \delta_k} = \limsup \theta_k \quad (3.39) \]
\[ = \sqrt{1 - \alpha^* \left[ 2(1 - \gamma t d^2) - \alpha^*(1 - (2\gamma t - 1)d^2) \right]} < 1, \]

for setting \( d = \sqrt{a^2/(e^2 + (2\gamma t - 1)a^2)}. \)
Algorithm 3.3 as

Suppose that the sequence \( \{x^k\} \) be a zero of Theorem 3.7. Let \( \eta : X \times X \to X \) be \((t)\)-strongly monotone. For an arbitrarily chosen initial point \( x^0 \), let the sequence \( \{x^k\} \) be bounded (in the sense that there exists at least one solution to \( 0 \in M(x) \)) and generated by Algorithm 3.3 as

\[
x^{k+1} = (1 - \alpha_k - \beta_k)x^k + \alpha_k y^k \quad \text{for } k \geq 0
\]

with

\[
\|y^k - J_{\xi_k}^{M, \eta}(x^k)\| \leq \epsilon_k,
\]

where \( J_{\xi_k}^{M, \eta} = (I + c_k M)^{-1} \), and sequences

\[
\{c_k\}, \{\alpha_k\}, \{\beta_k\} \subseteq [0, \infty)
\]

satisfy \( E_1 = \sum_{k=0}^{\infty} c_k < \infty \), \( \Delta_1 = \inf \alpha_k > 0 \), \( \Delta_2 = \sup \alpha_k < 2 \), and \( c^* = \inf c_k > 0 \).

In addition, one assumes (for \( \gamma > 0 \))

\[
\langle u - v, J_{\xi_k}^{M, \eta}(u) - J_{\xi_k}^{M, \eta}(v) \rangle \geq \gamma \langle u - v, \eta(J_{\xi_k}^{M, \eta}(u), J_{\xi_k}^{M, \eta}(v)) \rangle \quad \forall u, v \in X.
\]

Then the sequence \( \{x^k\} \) converges weakly to a solution of (1.1).

**Proof.** The proof is similar to that of the first part of Theorem 3.5 on applying the generalized representation lemma. \( \square \)

**Theorem 3.7.** Let \( X \) be a real Hilbert space. Let \( M : X \to 2^X \) be maximal \((\eta)\)-monotone, and let \( x^* \) be a zero of \( M \). Let \( \eta : X \times X \to X \) be \((t)\)-strongly monotone. Let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,
\]

and \( y^k \) satisfies

\[
\|y^k - J_{\xi_k}^{M, \eta}(x^k)\| \leq \epsilon_k,
\]

where \( J_{\xi_k}^{M, \eta} = (I + c_k M)^{-1} \), \( \sum_{k=0}^{\infty} c_k < \infty \), \( \{c_k\}, \{\epsilon_k\} \subseteq [0, \infty), c_k \not\to c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2t/(2t - 1) \).

Furthermore, assume (for \( \gamma > 0 \))

\[
\langle u - v, J_{\xi_k}^{M, \eta}(u) - J_{\xi_k}^{M, \eta}(v) \rangle \geq \gamma \langle u - v, \eta(J_{\xi_k}^{M, \eta}(u), J_{\xi_k}^{M, \eta}(v)) \rangle \quad \forall u, v \in X.
\]

Suppose that the sequence \( \{x^k\} \) is bounded in the sense that there exists at least one solution to \( 0 \in M(x) \).
Then (for \( t \geq 1 \))

\[
(2\gamma t - 1) \left\| J_{c_k}^{M,\eta}(x_k) - x^* \right\|^2 \leq \left\| x^k - x^* \right\|^2 - \left\| J_k(x_k) \right\|^2,
\]

where

\[
J_k^* = I - J_{c_k}^{M,\eta}.
\]

In addition, assume that the sequence \( \{x_k\} \) is generated by Algorithm 3.4 as well, and that \( M^{-1} \) is \((a)\)-Lipschitz continuous at 0, that is, there exists a unique solution \( z^* \) to \( 0 \in M(z) \) (equivalently, \( M^{-1}(0) = \{z^*\} \)) and for constants \( a \geq 0 \) and \( b > 0 \), one has

\[
\|z - z^*\| \leq a\|w\| \quad \text{whenever } z \in M^{-1}(w), \|w\| \leq b.
\]

Here

\[
\{\delta_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty)
\]

are scalar sequences such that \( \delta_k \to 0 \) and \( \sum_{k=0}^{\infty} \delta_k < \infty \).

Then the sequence \( \{x_k\} \) converges linearly to a unique solution \( x^* \) with rate

\[
\sqrt{1 - \alpha^* \left[ 2(1 - \gamma t d^2) - \alpha^* (1 - (2\gamma t - 1)d^2) \right]} < 1 \quad \text{for } t \geq 1,
\]

where \( d = \sqrt{a^2/(c^* + (2\gamma t - 1)a^2)} \), \( \alpha^* = \limsup_{k \to \infty} \alpha_k \), and sequences \( \{\alpha_k\} \) and \( \{c_k\} \) satisfy \( \alpha_k \geq 1 \), \( c_k \nearrow c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2\gamma t/(2\gamma t - 1) \).

Proof. The proof is similar to that of Theorem 3.5. \( \Box \)

4. Some Specializations

Finally, we examine some significant specializations of Theorem 3.5 in this section. Let us start with \( \gamma = 1 \) and \( t > 1 \) and applying Proposition 2.11.

**Theorem 4.1.** Let \( X \) be a real Hilbert space. Let \( M : X \to 2^X \) be maximal \((\eta)\)-monotone, and let \( x^* \) be a zero of \( M \). Let \( \eta : X \times X \to X \) be \((t)\)-strongly monotone. Furthermore, assume

\[
\langle u - v, J_{c_k}^{M,\eta}(u) - J_{c_k}^{M,\eta}(v) \rangle \geq \langle u - v, \eta(J_{c_k}^{M,\eta}(u), J_{c_k}^{M,\eta}(v)) \rangle \quad \forall u, v \in X.
\]

Let the sequence \( \{x_k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x_k + \alpha_k y_k \quad \forall k \geq 0,
\]
and \( y^k \) satisfies

\[
\left\| y^k - J^M_{c_k} (x^k) \right\| \leq e_k, \tag{4.3}
\]

where \( J^M_{c_k} = (I + c_k M)^{-1}, \sum_{k=0}^{\infty} e_k < \infty, \{\alpha_k\}, \{e_k\}, \{c_k\} \subseteq [0, \infty), c_k \not\rightarrow c^* \leq \infty, \inf_{k \geq 0} \alpha_k > 0, \) and \( \sup_{k \geq 0} \alpha_k < 2t/(2t - 1). \)

Suppose that the sequence \( \{x^k\} \) is bounded in the sense that there exists at least one solution to \( 0 \in M(x) \).

Then one has (for \( t > 1 \))

\[
(2t - 1) \left\| J^M_{c_k} (x^k) - x^* \right\|^2 \leq \left\| x^k - x^* \right\|^2 - \left\| J^*_k (x^k) \right\|^2, \tag{4.4}
\]

where \( t > 1 \) and

\[
J^*_k = I - J^M_{c_k}. \tag{4.5}
\]

In addition, suppose that the sequence \( \{x^k\} \) is generated by Algorithm 3.2 as well, and that \( M^{-1} \) is \( (a) \)-Lipschitz continuous at 0, that is, there exists a unique solution \( z^* \) to \( 0 \in M(z) \) (equivalently, \( M^{-1}(0) = \{z^*\} \)) and for constants \( a \geq 0 \) and \( b > 0 \), one has

\[
\|z - z^*\| \leq a\|w\| \quad \text{whenever} \quad z \in M^{-1}(w), \|w\| \leq b. \tag{4.6}
\]

Here

\[
\{\delta_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty) \tag{4.7}
\]

are scalar sequences such that \( \delta_k \rightarrow 0 \) and \( \sum_{k=0}^{\infty} \delta_k < \infty. \)

Then the sequence \( \{x^k\} \) converges linearly to a unique solution \( x^* \) with rate

\[
\sqrt{1 - \alpha^* [2(1-td^2) - \alpha^* (1-(2t-1)d^2)]} < 1 \quad \text{for} \ t > 1, \tag{4.8}
\]

where \( d = \sqrt{a^2/(c^*+2(2t-1)a^2)}, \alpha^* = \lim \sup_{k \rightarrow \infty} \alpha_k, \) and sequences \( \{\alpha_k\} \) and \( \{c_k\} \) satisfy \( \alpha_k \geq 1, c_k \not\rightarrow c^* \leq \infty, \inf_{k \geq 0} \alpha_k > 0, \) and \( \sup_{k \geq 0} \alpha_k < 2t/(2t - 1). \)

Proof. We need to include the proof for the sake of the completeness. Suppose that \( x^* \) is a zero of \( M. \) For all \( k \geq 0, \) we set

\[
J^*_k = I - J^M_{c_k}. \tag{4.9}
\]

Therefore, \( J^*_k (x^*) = 0. \) Then, in light of Theorem 3.1, any solution to (1.1) is a fixed point of \( J^{M_{c_k}} \), and hence a zero of \( J^*_k \).
Next, the proof of (4.4) follows from a regular manipulation, and the following equality:

$$\|u - v\|^2 = \left\| f^{M,\eta}_c(u) - f^{M,\eta}_c(v) + f'_k(u) - f'_k(v) \right\|^2 \quad \forall u, v \in X. \quad (4.10)$$

Before we start establishing linear convergence of the sequence \( \{x^k\} \), we express \( \{x^k\} \) in light of Algorithm 3.2 as

$$y^{k+1} = (1 - \alpha_k)x^k + \alpha_k f^{M,\eta}_c(x^k)$$

$$= (I - \alpha_k J'_k)(x^k). \quad (4.11)$$

Now we begin verifying the boundedness of the sequence \( \{x^k\} \) leading to \( x^k - f^{M,\eta}_c(x^k) \rightarrow 0 \).

Next, we estimate using Proposition 2.10 (for \( t \geq 1 \))

$$\left\| y^{k+1} - x^* \right\|^2 = \left\| (1 - \alpha_k)x^k + \alpha_k f^{M,\eta}_c(x^k) - x^* \right\|^2$$

$$= \left\| x^k - x^* - \alpha_k J'_k(x^k) \right\|^2$$

$$= \left\| x^k - x^* \right\|^2 - 2\alpha_k \left( x^k - x^* , J'_k(x^k) - J'_k(x^*) \right) + \alpha_k^2 \left\| J'_k(x^k) \right\|^2$$

$$\leq \left\| x^k - x^* \right\|^2 - \frac{2(t - 1)\alpha_k}{2t - 1} \left\| x^k - x^* \right\|^2 - \frac{2t}{2t - 1} \alpha_k^2 \left\| J'_k(x^k) \right\|^2$$

$$= \left( 1 - \frac{2(t - 1)\alpha_k}{2t - 1} \right) \left\| x^k - x^* \right\|^2 - \alpha_k \left( \frac{2t}{2t - 1} - \alpha_k \right) \left\| J'_k(x^k) \right\|^2. \quad (4.12)$$

Since under the assumptions \( \alpha_k (2t/(2t - 1) - \alpha_k) > 0 \), it follows that

$$\left\| y^{k+1} - x^* \right\| \leq \Delta \left\| x^k - x^* \right\| \leq \left\| x^k - x^* \right\|, \quad (4.13)$$

where \( \Delta = \sqrt{1 - 2(t - 1)\alpha_k/(2t - 1)} < 1 \).

Moreover,

$$\left\| x^{k+1} - y^{k+1} \right\| = \left\| (1 - \alpha_k)x^k + \alpha_k y^k - \left[ (1 - \alpha_k)x^k + \alpha_k f^{M,\eta}_c(x^k) \right] \right\|$$

$$= \left\| \alpha_k \left( y^k - J^{M,\eta}_c(x^k) \right) \right\|$$

$$\leq \alpha_k e_k. \quad (4.14)$$
Now we find the estimate leading to the boundedness of the sequence \( \{x^k\} \).

\[
\|x^{k+1} - x^*\| \leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\|
\leq \|x^k - x^*\| + \alpha_k e_k
\leq \|x^0 - x^*\| + \sum_{j=0}^k \alpha_j e_j
\leq \|x^0 - x^*\| + \frac{2t}{2t-1} \sum_{k=0}^\infty e_k.
\]

(4.15)

Thus, the sequence \( \{x^k\} \) is bounded.

We further examine the estimate

\[
\|x^{k+1} - x^*\|^2 = \|y^{k+1} - x^* + x^{k+1} - y^{k+1}\|^2
= \|y^{k+1} - x^*\|^2 + 2 \langle y^{k+1} - x^*, x^{k+1} - y^{k+1} \rangle + \|x^{k+1} - y^{k+1}\|^2
\leq \|y^{k+1} - x^*\|^2 + 2 \|y^{k+1} - x^*\| \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2
\leq \|x^k - x^*\|^2 - \alpha_k \left( \frac{2t}{2t-1} - \alpha_k \right) \|f_k(x^k)\|^2
+ 2 \left( \|x^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2 \right)
\leq \|x^k - x^*\|^2 - \alpha_k \left( \frac{2t}{2t-1} - \alpha_k \right) \|f_k(x^k)\|^2
+ 2 \left( \|x^0 - x^*\| + \frac{4t}{2t-1} \sum_{k=0}^\infty e_k \right) \frac{2t}{2t-1} \sum_{k=0}^\infty e_k + \left( \frac{2t}{2t-1} \right)^2 \sum_{k=0}^\infty e_k^2,
\]

(4.16)

where \( \alpha_k (2t/(2t-1) - \alpha_k) > 0 \).

Since \( \{e_k\} \) is summable, so is \( \{e_k^2\} \), and hence \( \sum_{k=0}^\infty e_k^2 < \infty \). As \( k \to \infty \), we have that

\[
\sum_{j=0}^k \|f_j(x^j)\|^2 < \infty \implies \lim_{k \to \infty} f_k(x^k) = 0,
\]

(4.17)

that is, \( x^k - f_{\alpha}^{M,q}(x^k) \to 0 \).
Now we turn our attention (using the previous argument) to linear convergence of the sequence \( \{x^k\} \). Since \( \lim_{k \to \infty} J_k^*(x^k) = 0 \), it implies for \( k \) large that \( c_k^{-1} J_k^*(x^k) \in M(J_k^{M,\eta}(x^k)) \). Moreover, \( \|c_k^{-1} J_k^*(x^k)\| \leq b \) for \( k \geq k' \) and \( b > 0 \). Therefore, in light of (4.6), by taking \( \omega = c_k^{-1} J_k^*(x^k) \) and \( z = J_k^{M,\eta}(x^k) \), we have

\[
\|J_k^{M,\eta}(x^k) - x^*\| \leq \alpha \|c_k^{-1} J_k^*(x^k)\| \quad \forall k \geq k'.
\]

Applying (4.4), we arrive at

\[
\|J_k^{M,\eta}(x^k) - x^*\|^2 \leq \frac{\alpha^2}{c_k^2 + (2t-1)a^2} \|x^k - x^*\|^2 \quad \text{for } t \geq 1,
\]

where \( J_k^{M,\eta}(x^*) = x^* \).

Since \( y^{k+1} := (1 - \alpha_k)x^k + \alpha_k J_k^{M,\eta}(x^k) \), we estimate using (4.1) and \( (\alpha_k \geq 1) \) that

\[
\|y^{k+1} - x^*\|^2
\]

\[
= \left\| (1 - \alpha_k)x^k + \alpha_k J_k^{M,\eta}(x^k) - x^* \right\|^2
\]

\[
= \|\alpha_k (J_k^{M,\eta}(x^k) - x^*) + (1 - \alpha_k)(x^k - x^*)\|^2
\]

\[
= \alpha_k^2 \|J_k^{M,\eta}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2 + 2\alpha_k(1 - \alpha_k) \langle J_k^{M,\eta}(x^k), x^k - x^* \rangle
\]

\[
\leq \alpha_k^2 \|J_k^{M,\eta}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2 + 2\alpha_k(1 - \alpha_k) \langle J_k^{M,\eta}(x^k), x^k - x^* \rangle
\]

\[
\leq \alpha_k^2 \|J_k^{M,\eta}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2 + 2\alpha_k(1 - \alpha_k)t \|J_k^{M,\eta}(x^k) - x^*\|^2
\]

\[
= \left[ 2\alpha_k(1 - \alpha_k)t + \alpha_k^2 \right] \|J_k^{M,\eta}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2
\]

\[
= \alpha_k [2t - (2t - 1)\alpha_k] \|J_k^{M,\eta}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2
\]

\[
\leq \alpha_k [2t - (2t - 1)\alpha_k] \frac{a^2}{c_k^2 + (2t-1)a^2} \|x^k - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2
\]

\[
= \left( \alpha_k [2t - (2t - 1)\alpha_k] \frac{a^2}{c_k^2 + (2t-1)a^2} + (1 - \alpha_k)^2 \right) \|x^k - x^*\|^2,
\]

(4.20)

where \( \alpha_k [2t - (2t - 1)\alpha_k] > 0 \).
Hence, we have

\[ \|y^{k+1} - x^*\| \leq \theta_k \|x^k - x^*\|, \tag{4.21} \]

where

\[ \theta_k = \sqrt{\alpha_k [2t - (2t - 1)\alpha_k] \left[ \frac{a^2}{c_k^2 + (2t - 1)a^2} \right] + (1 - \alpha_k)^2 < 1,} \tag{4.22} \]

for \( \alpha_k [2t - (2t - 1)\alpha_k] > 0 \) and \( \alpha_k \geq 1. \)

Since Algorithm 3.2 ensures

\[ \|y^k - J_{c_k^M \eta}^*(x^k)\| \leq \delta_k \|y^k - x^k\|, \]

\[ \alpha_k (y^k - x^k) = x^{k+1} - x^k, \tag{4.23} \]

we have

\[
\begin{align*}
\|x^{k+1} - y^{k+1}\| &= \|\alpha_k (y^k - J_{c_k^M \eta}^*(x^k))\| \\
&\leq \alpha_k \delta_k \|y^k - x^k\|
\end{align*}
\]

\[
\begin{align*}
\|x^{k+1} - x^*\| &= \|y^{k+1} - x^* + x^{k+1} - y^{k+1}\| \\
&\leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\| \\
&\leq \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^k\| \\
&\leq \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\| \\
&\leq \theta_k \|x^k - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\|.
\end{align*}
\]

It follows that

\[ \|x^{k+1} - x^*\| \leq \frac{\theta_k + \delta_k}{1 - \delta_k} \|x^k - x^*\|, \tag{4.25} \]

where

\[
\limsup \frac{\theta_k + \delta_k}{1 - \delta_k} = \limsup \theta_k
\]

\[
= \sqrt{1 - \alpha^* [2(1 - td^2) - \alpha^*(1 - (2t - 1)d^2)]} < 1, \tag{4.26}
\]

for setting \( d = \sqrt{a^2 / (c^2 + (2t - 1)a^2)}. \)
Second we examine Theorem 3.5 when $\gamma > 1$ and $t = 1$, but in this case there is no need to include the proof.

**Theorem 4.2.** Let $X$ be a real Hilbert space. Let $M : X \to 2^X$ be maximal $(\eta)$-monotone, and let $x^*$ be a zero of $M$. Let $\eta : X \times X \to X$ be strongly monotone. Furthermore, assume (for $\gamma > 1$)

$$\langle u - v, J_{c_k}^{M,\eta} (u) - J_{c_k}^{M,\eta} (v) \rangle \geq \gamma \langle u - v, \eta(J_{c_k}^{M,\eta} (u), J_{c_k}^{M,\eta} (v)) \rangle \quad \forall u, v \in X. \quad (4.27)$$

Let the sequence $\{x^k\}$ be generated by the iterative procedure

$$x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0, \quad (4.28)$$

and $y^k$ satisfies

$$\|y^k - J_{c_k}^{M,\eta} (x^k)\| \leq e_k, \quad (4.29)$$

where $J_{c_k}^{M,\eta} = (I + c_k M)^{-1}$, $\sum_{k=0}^{\infty} e_k < \infty$, $\{\alpha_k\}, \{e_k\} \subseteq [0, \infty)$, $c_k, c^* \leq \infty$, $\inf_{k \geq 0} \alpha_k > 0$, and $\sup_{k \geq 0} \alpha_k < 2\gamma / (2\gamma - 1)$.

Suppose that the sequence $\{x^k\}$ is bounded in the sense that there exists at least one solution to $0 \in M(x)$.

Then one has

$$\frac{2\gamma - 1}{2\gamma - 1} \|J_{c_k}^{M,\eta} (x^k) - x^*\|^2 \leq \|x^k - x^*\|^2 - \|J_k^* (x^k)\|^2, \quad (4.30)$$

where $\gamma > 1$ and

$$J_k^* = I - J_{c_k}^{M,\eta}. \quad (4.31)$$

In addition, suppose that the sequence $\{x^k\}$ is generated by Algorithm 3.2 as well, and that $M^{-1}$ is (a)-Lipschitz continuous at 0, that is, there exists a unique solution $z^*$ to $0 \in M(z)$ (equivalently, $M^{-1}(0) = \{z^*\}$) and for constants $a \geq 0$ and $b > 0$, one has

$$\|z - z^*\| \leq a\|w\| \quad \text{whenever } z \in M^{-1}(w), \|w\| \leq b. \quad (4.32)$$

Here

$$\{\delta_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty) \quad (4.33)$$

are scalar sequences such that $\delta_k \to 0$ and $\sum_{k=0}^{\infty} \delta_k < \infty$. 
Then the sequence \( \{x^k\} \) converges linearly to a unique solution \( x^* \) with rate

\[
\sqrt{1 - \alpha^* \left[ 2(1 - \gamma d^2) - \alpha^*(1 - (2\gamma - 1)d^2) \right]} < 1,
\]

where \( d = \sqrt{a^2 / (c^* + (2\gamma - 1)a^2)} \), \( \alpha^* = \limsup_{k \to \infty} \alpha_k \), and sequences \( \{\alpha_k\} \) and \( \{c_k\} \) satisfy \( \alpha_k \geq 1 \), \( c_k \nearrow c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2/(2\gamma - 1) \).

Finally, we consider the case when \( \gamma = t = 1 \) in Theorem 3.5, especially using Proposition 2.13. In this situation, the inclusion of the complete proof seems to be appropriate.

**Theorem 4.3.** Let \( X \) be a real Hilbert space. Let \( M : X \to 2^X \) be maximal \((\eta)\)-monotone, and let \( x^* \) be a zero of \( M \). Let \( \eta : X \times X \to X \) be strongly monotone. Furthermore, assume

\[
\langle u - v, J_{c_k}^{M,\eta}(u) - J_{c_k}^{M,\eta}(v) \rangle \geq \langle u - v, \eta(J_{c_k}^{M,\eta}(u), J_{c_k}^{M,\eta}(v)) \rangle \quad \forall u, v \in X.
\]

Let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,
\]

and \( y^k \) satisfies

\[
\|y^k - J_{c_k}^{M,\eta}(x^k)\| \leq e_k,
\]

where \( J_{c_k}^{M,\eta} = (I + c_k M)^{-1} \), \( \sum_{k=0}^\infty e_k < \infty \), \( \{\alpha_k\}, \{e_k\}, \{c_k\} \subseteq [0, \infty) \), \( c_k \nearrow c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2/2\gamma - 1 \).

Suppose that the sequence \( \{x^k\} \) is bounded in the sense that there exists at least one solution to \( 0 \in M(x) \).

Then one has

\[
\|J_{c_k}^{M,\eta}(x^k) - x^*\|^2 \leq \|x^k - x^*\|^2 - \|J_k^*(x^k)\|^2,
\]

where

\[
J_k^* = I - J_{c_k}^{M,\eta}.
\]

In addition, suppose that the sequence \( \{x^k\} \) is generated by Algorithm 3.2 as well, and that \( M^{-1} \) is \((a)\)-Lipschitz continuous at 0, that is, there exists a unique solution \( z^* \) to \( 0 \in M(z) \).
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(equivalently, \( M^{-1}(0) = \{ z^* \} \)) and for constants \( a \geq 0 \) and \( b > 0 \), one has

\[
\| z - z^* \| \leq a \| w \| \quad \text{whenever} \quad z \in M^{-1}(w), \quad \| w \| \leq b.
\]  \tag{4.40}

Here

\[
\{ \delta_k \}, \{ \alpha_k \}, \{ c_k \} \subseteq [0, \infty)
\]  \tag{4.41}

are scalar sequences such that \( \delta_k \to 0 \) and \( \sum_{k=0}^{\infty} \delta_k < \infty \).

Then the sequence \( \{ x^k \} \) converges linearly to a unique solution \( x^* \) with rate

\[
\sqrt{1 - \alpha^*[2(1 - d^2) - \alpha^*(1 - d^2)]} < 1,
\]  \tag{4.42}

where \( d = \sqrt{a^2/(c^*/a^2 + a^2)} \), \( \alpha^* = \limsup_{k \to \infty} \alpha_k \), and sequences \( \{ \alpha_k \} \) and \( \{ c_k \} \) satisfy \( \alpha_k \geq 1 \), \( c_k / c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2 \).

Proof. We need to include the proof for the sake of the completeness. Suppose that \( x^* \) is a zero of \( M \). For all \( k \geq 0 \), we set

\[
J^{\ast}_k = I - J_{M, \eta}^{c_k}.
\]  \tag{4.43}

Therefore, \( J^{\ast}_k(x^*) = 0 \). Then, in light of Theorem 3.1, any solution to (1.1) is a fixed point of \( J_{c_k}^{M, \eta} \), and hence a zero of \( J^{\ast}_k \).

Next, the proof of (4.38) follows from a regular manipulation, and the following equality:

\[
\| u - v \|^2 = \| J^{M, \eta} (u) - J^{M, \eta} (v) + J^*_k (u) - J^*_k (v) \|^2 \quad \forall u, v \in X.
\]  \tag{4.44}

Before we start establishing linear convergence of the sequence \( \{ x^k \} \), we express \( \{ x^k \} \) in light of Algorithm 3.2 as

\[
y^{k+1} = (1 - \alpha_k) x^k + \alpha_k J_{c_k}^{M, \eta} (x^k)
\]

\[
= (I - \alpha_k J^*_k)(x^k).
\]  \tag{4.45}

Now we begin examining the boundedness of the sequence \( \{ x^k \} \) leading to \( x^k - J_{c_k}^{M, \eta} (x^k) \to 0 \).
Next, we estimate using Proposition 2.13 that

$$\left\| y^{k+1} - x^* \right\|^2 = \left\| (1 - \alpha_k) x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) - x^* \right\|^2$$

$$= \left\| x^k - x^* - \alpha_k J_k^*(x^k) \right\|^2$$

$$= \left\| x^k - x^* \right\|^2 - 2\alpha_k \left( x^k - x^* - J_k^*(x^k) \right) + \alpha_k^2 \left\| J_k^*(x^k) \right\|^2$$

$$\leq \left\| x^k - x^* \right\|^2 - 2\alpha_k \left\| J_k^*(x^k) \right\|^2 + \alpha_k^2 \left\| J_k^*(x^k) \right\|^2$$

$$= \left\| x^k - x^* \right\|^2 - \alpha_k (2 - \alpha_k) \left\| J_k^*(x^k) \right\|^2.$$  \hspace{1cm} (4.46)

Since under the assumptions \( \alpha_k (2 - \alpha_k) > 0 \), it follows that

$$\left\| y^{k+1} - x^* \right\| \leq \left\| x^k - x^* \right\|. \hspace{1cm} (4.47)$$

Moreover,

$$\left\| x^{k+1} - y^{k+1} \right\| = \left\| (1 - \alpha_k) x^k + \alpha_k y^k - \left( (1 - \alpha_k) x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) \right) \right\|$$

$$= \left\| \alpha_k \left( y^k - J_{c_k}^{M,\eta}(x^k) \right) \right\|$$

$$\leq \alpha_k e_k.$$ \hspace{1cm} (4.48)

Now we find the estimate leading to the boundedness of the sequence \( \{ x^k \} \),

$$\left\| x^{k+1} - x^* \right\| \leq \left\| y^{k+1} - x^* \right\| + \left\| x^{k+1} - y^{k+1} \right\|$$

$$\leq \left\| x^k - x^* \right\| + \alpha_k e_k$$

$$\leq \left\| x^0 - x^* \right\| + \sum_{j=0}^{k} \alpha_j e_j$$ \hspace{1cm} (4.49)

$$\leq \left\| x^0 - x^* \right\| + 2 \sum_{k=0}^{\infty} \epsilon_k.$$ 

Thus, the sequence \( \{ x^k \} \) is bounded.
We further examine the estimate

\[ \| x^{k+1} - x^* \|^2 = \| y^{k+1} - x^* + x^{k+1} - y^{k+1} \|^2 \]

\[ = \| y^{k+1} - x^* \|^2 + 2 \langle y^{k+1} - x^*, x^{k+1} - y^{k+1} \rangle + \| x^{k+1} - y^{k+1} \|^2 \]

\[ \leq \| y^{k+1} - x^* \|^2 + 2 \| y^{k+1} - x^* \| \| x^{k+1} - y^{k+1} \| + \| x^{k+1} - y^{k+1} \|^2 \]

\[ \leq \| x^k - x^* \|^2 - \alpha_k (2 - \alpha_k) \| J_k^* (x^k) \|^2 \]

\[ + 2 \left( \| x^{k+1} - x^* \| + \| x^{k+1} - y^{k+1} \| \right) \| x^{k+1} - y^{k+1} \| + \| x^{k+1} - y^{k+1} \|^2 \]

\[ \leq \| x^k - x^* \|^2 - \alpha_k (2 - \alpha_k) \| J_k^* (x^k) \|^2 + 2 \left( \| x^0 - x^* \| + \sum_{k=0}^{\infty} e_k \right) 2 \sum_{k=0}^{\infty} e_k + 4 \sum_{k=0}^{\infty} e_k^2, \]

where \( \alpha_k (2 - \alpha_k) > 0 \).

Since \( \{ e_k \} \) is summable, so is \( \{ e_k^2 \} \), and hence \( \sum_{k=0}^{\infty} e_k^2 < \infty \). As \( k \to \infty \), we have that

\[ \sum_{j=0}^{k} \| J_j^* (x^j) \|^2 < \infty \iff \lim_{k \to \infty} J_k^* (x^k) = 0, \]

(4.51)

that is, \( x^k - J^M_{\alpha_k} (x^k) \to 0 \). Now we find the estimate leading to the boundedness of the sequence \( \{ x^k \} \).

\[ \| x^{k+1} - x^* \| \leq \| y^{k+1} - x^* \| + \| x^{k+1} - y^{k+1} \| \]

\[ \leq \| x^k - x^* \| + \alpha_k e_k \]

\[ \leq \| x^0 - x^* \| + \sum_{j=0}^{k} \alpha_j e_j \]

\[ \leq \| x^0 - x^* \| + \sum_{k=0}^{\infty} e_k. \]

(4.52)

Thus, the sequence \( \{ x^k \} \) is bounded.
We further examine the estimate

\[
\|x^{k+1} - x^*\|^2 = \|y^{k+1} - x^{k+1} - y^{k+1}\|^2
\]

\[
= \|y^{k+1} - x^*\|^2 + 2 \langle y^{k+1} - x^*, x^{k+1} - y^{k+1} \rangle + \|x^{k+1} - y^{k+1}\|^2
\]

\[
\leq \|y^{k+1} - x^*\|^2 + 2 \|y^{k+1} - x^*\| \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2
\]

\[
\leq \|x^k - x^*\|^2 - a_k(2 - a_k) \|J_k'(x^k)\|^2 + 2 \left( \|x^{k+1} - y^{k+1}\| \|x^{k+1} - y^{k+1}\| + \|x^{k+1} - y^{k+1}\|^2 \right)
\]

\[
\leq \|x^k - x^*\|^2 - a_k(2 - a_k) \|J_k'(x^k)\|^2 + 2 \left( \|x^0 - x^*\| + \|x^0 - x^*\| + 4 \sum_{k=0}^{\infty} e_k \right) \|x^0 - x^*\| + 4 \sum_{k=0}^{\infty} e_k + 4 \sum_{k=0}^{\infty} e_k^2
\]

(4.53)

where \(a_k(2 - a_k) > 0\).

Since \(\{e_k\}\) is summable, so is \(\{e_k^2\}\), and hence \(\sum_{k=0}^{\infty} e_k^2 < \infty\). As \(k \to \infty\), we have that

\[
\sum_{j=0}^{k} \|J_j'(x^j)\|^2 < \infty \implies \lim_{k \to \infty} J_k'(x^k) = 0,
\]

(4.54)

that is, \(x^k - J_{c_k}^{M,\eta}(x^k) \to 0\).

Now we turn our attention (using the previous argument) to linear convergence of the sequence \(\{x^k\}\). Since \(\lim_{k \to \infty} J_k'(x^k) = 0\), it implies for \(k\) large that \(c_k^{-1} J_k'(x^k) \in M(J_{c_k}^{M,\eta}(x^k))\). Moreover, \(\|c_k^{-1} J_k'(x^k)\| \leq b\) for \(k \geq k'\) and \(b > 0\). Therefore, in light of (4.40), by taking \(\omega = c_k^{-1} J_k'(x^k)\) and \(z = J_{c_k}^{M,\eta}(x^k)\), we have

\[
\|J_{c_k}^{M,\eta}(x^k) - x^*\| \leq a \|c_k^{-1} J_k'(x^k)\| \quad \forall k \geq k'.
\]

(4.55)

Applying (4.38), we arrive at

\[
\|J_{c_k}^{M,\eta}(x^k) - x^*\|^2 \leq \frac{\alpha^2}{c_k + d^2} \|x^k - x^*\|^2 \quad \text{for} \ t \geq 1,
\]

(4.56)

where \(J_{c_k}^{M,\eta}(x^*) = x^*\).
Since \( y^{k+1} := (1 - \alpha_k)x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) \), we estimate using (4.35) and \((\alpha_k \geq 1)\) that

\[
\| y^{k+1} - x^* \|^2 \\
= \| (1 - \alpha_k)x^k + \alpha_k J_{c_k}^{M,\eta}(x^k) - x^* \|^2 \\
= \| \alpha_k \left( J_{c_k}^{M,\eta}(x^k) - x^* \right) + (1 - \alpha_k)(x^k - x^*) \|^2 \\
= \alpha_k^2 \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 + 2\alpha_k (1 - \alpha_k) \left( J_{c_k}^{M,\eta}(x^k) - x^*, x^k - x^* \right) \\
\leq \alpha_k^2 \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 + 2\alpha_k (1 - \alpha_k) \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 \\
= \left[ 2\alpha_k (1 - \alpha_k) + \alpha_k^2 \right] \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 \\
= \alpha_k (2 - \alpha_k) \| J_{c_k}^{M,\eta}(x^k) - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 \\
\leq \alpha_k (2 - \alpha_k) \frac{\alpha^2}{c_k^2 + a^2} \| x^k - x^* \|^2 + (1 - \alpha_k)^2 \| x^k - x^* \|^2 \\
= \alpha_k (2 - \alpha_k) \frac{\alpha^2}{c_k^2 + a^2} + (1 - \alpha_k)^2 \| x^k - x^* \|^2. \\
\] (4.57)

Hence, we have

\[
\| y^{k+1} - x^* \| \leq \theta_k \| x^k - x^* \|, \\
\] (4.58)

where

\[
\theta_k = \sqrt{\alpha_k (2 - \alpha_k) \left( \frac{\alpha^2}{c_k^2 + a^2} \right) + (1 - \alpha_k)^2} < 1, \\
\] (4.59)

for \( \alpha_k [2 - \alpha_k] > 0 \) and \( \alpha_k \geq 1 \).

Since Algorithm 3.2 ensures

\[
\| y^k - J_{c_k}^{M,\eta}(x^k) \| \leq \delta_k \| y^k - x^k \|, \]

\[
\alpha_k \left( y^k - x^k \right) = x^{k+1} - x^k, \\
\] (4.60)
we have
\[
\|x^{k+1} - y^{k+1}\| = \|x^k - f_{\alpha_k}^M(x^k)\| \leq \alpha_k \|y^k - x^k\|,
\]
\[
\|x^{k+1} - x^*\| = \|y^{k+1} - x^* + x^{k+1} - y^{k+1}\|
\leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\|
\leq \|y^{k+1} - x^*\| + \alpha_k \delta_k \|y^k - x^k\|
= \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^k\|
\leq \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\|
\leq \theta_k \|x^k - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\|.
\]

It follows that
\[
\|x^{k+1} - x^*\| \leq \frac{\theta_k + \delta_k}{1 - \delta_k} \|x^k - x^*\|,
\]
where
\[
\limsup \frac{\theta_k + \delta_k}{1 - \delta_k} = \limsup \theta_k
= \sqrt{1 - \alpha^* \left[2(1 - d^2) - \alpha^* (1 - d^2)\right]} < 1,
\]
for setting \(d = \sqrt{a^2/(c^2 + a^2)}\).

Note that if we set \(\eta(x, y) = x - y\) in Theorem 4.3, we get a result connecting [2] to the case of a linear convergence setting, but the algorithm remains overrelaxed (or superrelaxed). In this context, we state the following results before we start examining Theorem 4.7, the main result on linear convergence in the maximal monotone setting. Note that based on Proposition 4.6, notions of cocoercivity and firm nonexpansiveness coincide, though it is well known that they may differ in usage much depending on the context.

**Theorem 4.4.** Let \(X\) be a real Hilbert space, and let \(M : X \to 2^X\) be maximal monotone. Then the following statements are mutually equivalent.

(i) An element \(u \in X\) is a solution to (1.1).

(ii) For an \(u \in X\), one has
\[
u = J_c^M(u) \quad \text{for } c > 0,
\]
where
\[
J_c^M(u) = (I + cM)^{-1}(u).
\]
Proof. It follows from the definition of the generalized resolvent operator corresponding to \( M \).

Next, we present the super-relaxed Proximal point algorithm based on the maximal monotonicity.

Algorithm 4.5. Let \( M : X \to 2^X \) be a set-valued maximal monotone mapping on \( X \) with \( 0 \in \text{range}(M) \), and let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,
\]

and \( y^k \) satisfies

\[
\|y^k - J^M_{c_k}(x^k)\| \leq \delta_k \|y^k - x^k\|,
\]

where \( J^M_{c_k} = (I + c_k M)^{-1} \), \( \delta_k \to 0 \) and

\[
y^{k+1} = (1 - \alpha_k)x^k + \alpha_k J^M_{c_k}(x^k) \quad \forall k \geq 0.
\]

Here

\[
\{\delta_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty)
\]

are scalar sequences such that \( \sum_{k=0}^{\infty} \delta_k < \infty \).

Proposition 4.6. Let \( X \) be a real Hilbert space, and let \( M : X \to 2^X \) be maximal monotone. Then, for \( J^*_k = I - J^M_k \), one has

\[
\langle u - v, J^*_k(u) - J^*_k(v) \rangle \geq \|J^*_k(u) - J^*_k(v)\|^2,
\]

where

\[
J^M_k(u) = (I + \rho M)^{-1}(u) \quad \forall u \in X.
\]

Theorem 4.7. Let \( X \) be a real Hilbert space. Let \( M : X \to 2^X \) be maximal monotone, and let \( x^* \) be a zero of \( M \). Let the sequence \( \{x^k\} \) be generated by the iterative procedure

\[
x^{k+1} = (1 - \alpha_k)x^k + \alpha_k y^k \quad \forall k \geq 0,
\]

and \( y^k \) satisfies

\[
\|y^k - J^M_{c_k}(x^k)\| \leq e_k,
\]
where \( J^M_k = (I + c_k M)^{-1} \), \( \sum_{k=0}^{\infty} e_k < \infty \), \( |\alpha_k|, |e_k| |c_k| \subseteq [0, \infty) \), \( c_k \nearrow c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2 \).

Suppose that the sequence \( \{x^k\} \) is bounded in the sense that there exists at least one solution to \( 0 \in M(x) \).

Then one has

\[
\left\| J^M_c(x^k) - x^* \right\|^2 \leq \left\| x^k - x^* \right\|^2 - \left\| J^*_{c_k}(x^k) \right\|^2,
\]

(4.74)

where \( J^*_{c_k} = I - J^M_{c_k} \).

In addition, suppose that the sequence \( \{x^k\} \) is generated by Algorithm 4.5, and that \( M^{-1} \) is \((\alpha)\)-Lipschitz continuous at 0, that is, there exists a unique solution \( z^* \) to \( 0 \in M(z) \) (equivalently, \( M^{-1}(0) = \{z^*\} \)) and for constants \( a \geq 0 \) and \( b > 0 \), one has

\[
\|z - z^*\| \leq a\|\omega\| \quad \text{whenever} \quad z \in M^{-1}(\omega), \quad \|\omega\| \leq b.
\]

Here

\[
\{\delta_k\}, \{\alpha_k\}, \{c_k\} \subseteq [0, \infty)
\]

(4.77)

are scalar sequences such that \( \delta_k \to 0 \) and \( \sum_{k=0}^{\infty} \delta_k < \infty \).

Then the sequence \( \{x^k\} \) converges linearly to a unique solution \( x^* \) with rate

\[
\sqrt{1 - \alpha^* [2(1 - d^2) - \alpha^*(1 - d^2)]} < 1,
\]

(4.78)

where \( d = \sqrt{a^2/(c^* - \alpha)} \), \( \alpha^* = \lim \sup_{k \to \infty} \alpha_k \), and sequences \( \{\alpha_k\} \) and \( \{c_k\} \) satisfy \( \alpha_k \geq 1 \), \( c_k \nearrow c^* \leq \infty \), \( \inf_{k \geq 0} \alpha_k > 0 \), and \( \sup_{k \geq 0} \alpha_k < 2 \).

Proof. We need to include the proof for the sake of the completeness. Suppose that \( x^* \) is a zero of \( M \). For all \( k \geq 0 \), we set

\[
J^*_{c_k} = I - J^M_{c_k}.
\]

(4.79)

Therefore, \( J^*_{c_k}(x^*) = 0 \). Then, in light of Theorem 4.4, any solution to (1.1) is a fixed point of \( J^M_{c_k} \), and hence a zero of \( J^*_{c_k} \).

Next, the proof of (4.74) follows from applying the regular manipulation, and the following equality:

\[
\|u - v\|^2 = \left\| J^M_{c_k}(u) - J^M_{c_k}(v) + J^*_{c_k}(u) - J^*_{c_k}(v) \right\|^2 \quad \forall u, v \in X.
\]

(4.80)

Before we start establishing linear convergence of the sequence \( \{x^k\} \), we express \( \{x^k\} \) in light
of Algorithm 4.5 as
\[ y^{k+1} = (1 - a_k)x^k + a_k f^{M}_{c_k}(x^k) \]
\[ = (I - a_k f^*_k)\left(x^k\right). \] (4.81)

Now we begin examining the boundedness of the sequence \(\{x^k\}\) leading to \(x^k - f^{M}_{c_k}(x^k) \to 0\).

Next, we estimate using Proposition 4.6 that
\[ \|y^{k+1} - x^*\|^2 = \|(1 - a_k)x^k + a_k f^{M}_{c_k}(x^k) - x^*\|^2 \]
\[ = \|x^k - x^* - a_k f^*_k(x^k)\|^2 \]
\[ = \|x^k - x^*\|^2 - 2a_k \left(x^k - x^*, f^*_k(x^k) - f^*_k(x^*)\right) + a_k^2 \|f^*_k(x^k)\|^2 \]
\[ \leq \|x^k - x^*\|^2 - 2a_k \|f^*_k(x^k)\|^2 + a_k^2 \|f^*_k(x^k)\|^2 \]
\[ = \|x^k - x^*\|^2 - a_k (2 - a_k) \|f^*_k(x^k)\|^2. \] (4.82)

Since under the assumptions \(a_k (2 - a_k) > 0\), it follows that
\[ \|y^{k+1} - x^*\| \leq \|x^k - x^*\|. \] (4.83)

Moreover,
\[ \|x^{k+1} - y^{k+1}\| = \|(1 - a_k)x^k + a_k y^k - \left((1 - a_k)x^k + a_k f^{M}_{c_k}(x^k)\right)\| \]
\[ = \|a_k (y^k - f^{M}_{c_k}(x^k))\| \]
\[ \leq a_k e_k. \] (4.84)

Now we find the estimate leading to the boundedness of the sequence \(\{x^k\}\),
\[ \|x^{k+1} - x^*\| \leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\| \]
\[ \leq \|x^k - x^*\| + a_k e_k \]
\[ \leq \|x^0 - x^*\| + \sum_{j=0}^{k} a_j e_j \]
\[ \leq \|x^0 - x^*\| + \sum_{k=0}^{\infty} a_k e_k. \] (4.85)

Therefore, the sequence \(\{x^k\}\) is bounded.
We further examine the estimate

\[
\|x_{k+1} - x^*\|^2 = \|y_{k+1} - x^* + x_{k+1} - y_{k+1}\|^2
\]

\[
= \|y_{k+1} - x^*\|^2 + 2\langle y_{k+1} - x^*, x_{k+1} - y_{k+1}\rangle + \|x_{k+1} - y_{k+1}\|^2
\]

\[
\leq \|y_{k+1} - x^*\|^2 + 2\|y_{k+1} - x^*\|\|x_{k+1} - y_{k+1}\| + \|x_{k+1} - y_{k+1}\|^2
\]

\[
\leq \|x_k - x^*\|^2 - \alpha_k (2 - \alpha_k) \|J'_k(x^*)\|^2
\]

\[
+ 2 \left( \|x_{k+1} - x^*\| + \|x_{k+1} - y_{k+1}\| \right) \|x_{k+1} - y_{k+1}\| + \|x_{k+1} - y_{k+1}\|^2
\]

\[
\leq \|x_k - x^*\|^2 - \alpha_k (2 - \alpha_k) \|J'_k(x^*)\|^2 + 2 \left( \|x^0 - x^*\| + \sum_{k=0}^{\infty} e_k \right) \sum_{k=0}^{\infty} e_k + 4 \sum_{k=0}^{\infty} e_k^2,
\]

where \(\alpha_k (2 - \alpha_k) > 0\).

Since \(\{e_k\}\) is summable, so is \(\{e_k^2\}\), and hence \(\sum_{k=0}^{\infty} e_k^2 < \infty\). As \(k \to \infty\), we have that

\[
\sum_{j=0}^{k} \|J'_j(x^j)\|^2 < \infty \implies \lim_{k \to \infty} J'_k(x^k) = 0,
\]

that is, \(x^k - J_{\alpha_k}^{M\eta}(x^k) \to 0\). Now we find the estimate leading to the boundedness of the sequence \(\{x^k\}\),

\[
\|x_{k+1} - x^*\| \leq \|y_{k+1} - x^*\| + \|x_{k+1} - y_{k+1}\|
\]

\[
\leq \|x_k - x^*\| + \alpha_k e_k
\]

\[
\leq \|x^0 - x^*\| + \sum_{j=0}^{k} \alpha_j e_j
\]

\[
\leq \|x^0 - x^*\| + 2 \sum_{k=0}^{\infty} e_k.
\]

Thus, the sequence \(\{x^k\}\) is bounded.
We further examine the estimate

\[
\left\| x^{k+1} - x^* \right\|^2 = \left\| y^{k+1} - x^* + x^{k+1} - y^{k+1} \right\|^2 \\
= \left\| y^{k+1} - x^* \right\|^2 + 2 \left( y^{k+1} - x^*, x^{k+1} - y^{k+1} \right) + \left\| x^{k+1} - y^{k+1} \right\|^2 \\
\leq \left\| y^{k+1} - x^* \right\|^2 + 2 \left\| y^{k+1} - x^* \right\| \left\| x^{k+1} - y^{k+1} \right\| + \left\| x^{k+1} - y^{k+1} \right\|^2 \\
\leq \left\| x^k - x^* \right\|^2 - \alpha_k (2 - \alpha_k) \left\| J^*_k(x^k) \right\|^2 \\
+ 2 \left( \left\| x^{k+1} - x^* \right\| + \left\| x^{k+1} - y^{k+1} \right\| \right) \left\| x^{k+1} - y^{k+1} \right\| + \left\| x^{k+1} - y^{k+1} \right\|^2 \\
\leq \left\| x^k - x^* \right\|^2 - \alpha_k (2 - \alpha_k) \left\| J^*_k(x^k) \right\|^2 + 2 \left( \left\| x^0 - x^* \right\| + 4 \sum_{k=0}^\infty e_k \right) 2 \sum_{k=0}^\infty e_k + 4 \sum_{k=0}^\infty e_k^2, \\
(4.89)
\]

where \( \alpha_k (2 - \alpha_k) > 0 \).

Since \( \{ e_k \} \) is summable, so is \( \{ e_k^2 \} \), and hence \( \sum_{k=0}^\infty e_k^2 < \infty \). As \( k \rightarrow \infty \), we have that

\[
\sum_{j=0}^k \left\| J^*_j(x^j) \right\|^2 < \infty \iff \lim_{k \rightarrow \infty} J^*_k(x^k) = 0, \\
(4.90)
\]

that is, \( x^k - J^M_{c_k}(x^k) \rightarrow 0 \).

Now we turn our attention (using the previous argument) to linear convergence of the sequence \( \{ x^k \} \). Since \( \lim_{k \rightarrow \infty} J^*_k(x^k) = 0 \), it implies for \( k \) large that \( c_k^{-1} J^*_k(x^k) \in M(J^M_{c_k}(x^k)) \).

Moreover, \( \| c_k^{-1} J^*_k(x^k) \| \leq b \) for \( k \geq k' \) and \( b > 0 \). Therefore, in light of (4.76), by taking \( w = c_k^{-1} J^*_k(x^k) \) and \( z = J^M_{c_k}(x^k) \), we have

\[
\left\| J^M_{c_k}(x^k) - x^* \right\| \leq a \left\| c_k^{-1} J^*_k(x^k) \right\| \forall k \geq k'. \\
(4.91)
\]

Applying (4.74), we arrive at

\[
\left\| J^M_{c_k}(x^k) - x^* \right\|^2 \leq \frac{a^2}{c_k^2 + a^2} \left\| x^k - x^* \right\|^2 \text{ for } t \geq 1, \\
(4.92)
\]

where \( J^M_{c_k}(x^*) = x^* \).
Since $y^{k+1} := (1 - \alpha_k)x^k + \alpha_k J^M_{c_k}(x^k)$, we estimate (for $\alpha_k \geq 1$) that

$$
\|y^{k+1} - x^*\|^2 = \|(1 - \alpha_k)x^k + \alpha_k J^M_{c_k}(x^k) - x^*\|^2
$$

$$
= \|\alpha_k \left(J^M_{c_k}(x^k) - x^*\right) + (1 - \alpha_k)(x^k - x^*)\|^2
$$

$$
= \alpha_k^2 \|J^M_{c_k}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2 + 2\alpha_k(1 - \alpha_k) \left\langle J^M_{c_k}(x^k) - x^*, x^k - x^* \right\rangle
$$

$$
\leq \alpha_k^2 \|J^M_{c_k}(x^k) - x^*\|^2 + 2\alpha_k(1 - \alpha_k) \left\langle J^M_{c_k}(x^k) - x^*, x^k - x^* \right\rangle
$$

$$
= \left[2\alpha_k(1 - \alpha_k) + \alpha_k^2\right] \|J^M_{c_k}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2
$$

$$
\leq \alpha_k(2 - \alpha_k) \|J^M_{c_k}(x^k) - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2
$$

$$
\leq \alpha_k(2 - \alpha_k) \left\| \frac{a^2}{c_k^2 + a^2} \|x^k - x^*\|^2 + (1 - \alpha_k)^2 \|x^k - x^*\|^2 \right\|
$$

$$
= \alpha_k(2 - \alpha_k) \left\| \frac{a^2}{c_k^2 + a^2} + (1 - \alpha_k)^2 \right\| \|x^k - x^*\|^2.
$$

(4.93)

Hence, we have

$$
\|y^{k+1} - x^*\| \leq \theta_k \|x^k - x^*\|,
$$

(4.94)

where

$$
\theta_k = \sqrt{\alpha_k(2 - \alpha_k) \left\{ \frac{a^2}{c_k^2 + a^2} + (1 - \alpha_k)^2 \right\} < 1},
$$

(4.95)

for $\alpha_k \{2 - \alpha_k\} > 0$ and $\alpha_k \geq 1$.

Since Algorithm 4.5 ensures

$$
\|y^k - J^M_{c_k}(x^k)\| \leq \delta_k \|y^k - x^k\|,
$$

$$
\alpha_k(y^k - x^k) = x^{k+1} - x^k,
$$

(4.96)
we have
\[
\|x^{k+1} - y^{k+1}\| = \|\alpha_k (y^k - f^{M_n}_{c_k}(x^k))\| \leq \alpha_k \delta_k \|y^k - x^k\|,
\]
\[
\|x^{k+1} - x^*\| = \|y^{k+1} - x^* + x^{k+1} - y^{k+1}\|
\]
\[
\leq \|y^{k+1} - x^*\| + \|x^{k+1} - y^{k+1}\|
\leq \|y^{k+1} - x^*\| + \alpha_k \delta_k \|y^k - x^k\|
\leq \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^k\|
\]
\[
\leq \|y^{k+1} - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\|
\leq \theta_k \|x^k - x^*\| + \delta_k \|x^{k+1} - x^*\| + \delta_k \|x^k - x^*\|.
\]
It follows that
\[
\|x^{k+1} - x^*\| \leq \frac{\theta_k + \delta_k}{1 - \delta_k} \|x^k - x^*\|,
\]  
(4.98)

where
\[
\limsup \frac{\theta_k + \delta_k}{1 - \delta_k} = \limsup \theta_k
\]  
(4.99)

\[
= \sqrt{1 - \alpha^*(2 - \alpha^*)(1 - d^2)} < 1,
\]

for setting \( d = \sqrt{\frac{\alpha^2}{(c^2 + \alpha^2)}} \).

**References**


