Contents lists available at ScienceDirect

Computers and Mathematics with Applications

journal homepage: www.elsevier.com/locate/camwa

Numerical approach to monotone variational inequalities by a one-step projected reflected gradient method with line-search procedure

ABSTRACT

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ARTICLE INFO

Article history: Received 5 January 2016 Received in revised form 6 April 2016 Accepted 22 May 2016 Available online 1 July 2016

Keywords: Variational inequality Projection method Monotone operator Projected reflected method

1. Introduction

Throughout this paper, *C* is a closed convex set of \mathbb{R}^N , while $\langle ., . \rangle$ and $| \cdot |$ denote the usual inner product of \mathbb{R}^N and its induced norm, respectively. We aim at revisiting the computation of an element of the solution set, denoted by *S*, of the classical variational inequality problem:

find
$$u \in C$$
 such that $\langle F(u), v - u \rangle \ge 0 \quad \forall v \in C$, (1.1)

projection step onto the feasible set at each iteration.

In this paper we revisit a recent approach to classical monotone variational inequalities

by means of a projected reflected gradient-type method in \mathbb{R}^N . A line-search procedure is

incorporated for possible varying step-sizes and no Lipschitz-continuity condition on the

operator is required. A main feature of the proposed method is that it requires only one

where $F : \mathbb{R}^N \to \mathbb{R}^N$ is a given mapping such that:

$S \neq \emptyset;$	(1.2a)
<i>F</i> is monotone over \mathbb{R}^N , i.e., for all $x, y \in \mathbb{R}^N$: $\langle F(x) - F(y), x - y \rangle \ge 0$;	(1.2b)

F is continuous over \mathbb{R}^N .

This formalism (first introduced by Stampacchia in [1]) is well-known to provide a unified framework for the study of many significant real-world problems arising in mechanics, economics and so on (see, e.g., [2-5] and the references therein). It is worthwhile underlining that (1.1) gave rise to many algorithmic solutions depending on the structure of *C* and the properties of *F*. In particular when *F* and *C* do not possess any special structure the proposed numerical approaches to (1.1) are mainly based upon projection techniques onto *C*; see, e.g., [6].

Definition 1.1. The metric projection $P_C : \mathbb{R}^N \to C$ is the operator defined for all $x \in \mathbb{R}^N$ by $P_C x := \operatorname{argmin}_{z \in C} |z - x|$.







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(1.2c)

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http://dx.doi.org/10.1016/j.camwa.2016.05.028 0898-1221/© 2016 Elsevier Ltd. All rights reserved.

The use of such techniques was suggested by the following equivalent fixed point formulation of (1.1): find $u \in C$ such that $u = P_C(u - \lambda F(u))$, where λ is any positive real number.

Let us recall that the oldest method for solving (1.1) in a general framework is the following so-called extra-gradient method (introduced by Korpelevich [7]):

$$\overline{x}_n = P_C(x_n - \lambda_n F(x_n)), \qquad x_{n+1} = P_C(x_n - \lambda_n F(\overline{x}_n)), \tag{1.3}$$

where the step-sizes (λ_n) are positive real numbers. This algorithm involves two projection steps per iteration and its convergence was initially established under monotonicity and Lipschitz continuity of *F* for constant step-sizes (depending on the Lipschitz constant of *F*). Later on, the extra-gradient method has been enhanced through several extensions involving Armijo-type rules (see, e.g., Khobotov [8], Marcotte [9], Sun [10], Iusem [11], Tseng [12]) and outer approximation techniques (see, e.g., Solodov and Svaiter [13]). These latter strategies are specifically well-adapted to the situation when *F* is not Lipschitz continuous or no estimate of the Lipschitz constant is available.

This paper is mostly concerned with the case when the evaluation of the projection operator onto *C* is computationally expensive. A crucial feature regarding the design of numerical methods related to such a context is to minimize the number of evaluation of P_C per iteration. As famous algorithms with such interesting features we mention that ones discussed by lusem–Svaiter [14] and Solodov–Svaiter [13]. These methods (that involve a more effective line-search procedure) were able to drop the Lipschitz continuity condition even for a pseudo-monotone mapping *F* (also see lusem–Pérez [15] for extension to nonsmooth cases of *F*). These latter algorithms involve one projection onto *C* at each iteration together with another projection is needed in the line-search procedure. Modified extra-gradient methods with only one projection onto *C* per iteration has been also investigated. As an example, a special case (relative to (1.1)) of the general method proposed by Tseng [16] formally involves one projection step but its convergence was established by using an Armijo–Goldstein-type stepsize rule for which the trial values of step-sizes require additional evaluations of P_C . Other examples of such methods combine one projection onto *C* per iteration together with a cheaper projection step onto some hyperplane (see, e.g., Censor, Gibali and Reich [17], Malitsky and Semenov [18]). However, their convergence was stated under the condition of Lipschitz continuity of *F* for step-size rules that depend on the Lipschitz constant.

In this work we focus our attention on an alternative approach to (1.1) based on the following so-called projected reflected gradient method recently proposed by Malitsky [19]:

$$y_n = 2x_n - x_{n-1}, \qquad x_{n+1} = P_C(x_n - \lambda_n F(y_n)),$$
(1.4)

with positive step-sizes (λ_n) . This latter algorithm (formally) involves only one evaluation for each of the operators P_C and F per iteration. It was investigated in the context of a Hilbert space together with conditions of monotonicity and L-Lipschitz continuity of F (for some positive value L) over the whole space. The convergence of (1.4) was obtained is the special case of a constant step-size $\lambda \in (0, \frac{\sqrt{2}-1}{L})$. A convergence result was also established for varying step-sizes (λ_n) given by a specific procedure that (unfortunately) involves the computation of additional projections onto C. Nonetheless numerical experiments have been performed in [19] showing interesting and promising features (regarding the computational viewpoint) comparing with other numerical strategies.

The following variant of (1.4) has been also investigated (in a Hilbert space under the same conditions on *F*) in [20]:

$$y_n = x_n + \theta_n(x_n - x_{n-1}), \qquad x_{n+1} = P_C(x_n - \lambda_n F(y_n)),$$
(1.5)

where $\theta_n = \frac{\lambda_n}{\delta \lambda_{n-1}}$ for some $\delta \in (0, 1]$. The convergence of (1.5) was stated under the general conditions below on the step-sizes:

$$\lambda_n |F(y_n) - F(y_{n-1})| \le \epsilon \delta(\sqrt{2} - 1) |y_n - y_{n-1}|, \quad \text{for some } \epsilon \in (0, 1),$$
(1.6a)

$$\lambda_n \le \lambda_{n-1} \left(\delta + \frac{\lambda_{n-1}}{\lambda_{n-2}} \right)^{1/2},\tag{1.6b}$$

 $(\lambda_n) \subset [\bar{\mu}, \bar{\nu}]$ for some positive $\bar{\mu}$ and $\bar{\nu}$.

Our purpose here is to propose an enhanced variant of (1.5) for solving (1.1) under condition (1.2) (so we drop the Lipschitz continuity of the operator F) by incorporating a linesearch procedure that does not require any additional evaluation of P_c .

2. The algorithm and its convergence

For the sake of simplicity, we sometimes use the following notations: $\dot{x}_n = x_n - x_{n-1}$ and $\dot{y}_n = y_n - y_{n-1}$. In order to compute a solution of (1.1) we focus on the sequence (x_n) generated by the following method.

(1.6c)

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