# FINDING A ZERO OF THE SUM OF TWO MAXIMAL MONOTONE OPERATORS WITH MINIMIZATION PROBLEM 

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#### Abstract

The aim of this paper is to construct a new method for finding the zeros of the sum of two maximally monotone mappings in Hilbert spaces. We will define a simple function such that its set of zeros coincide with that of the sum of two maximal monotone operators. Moreover, we will use the Newton-Raphson algorithm to get an approximate zero. In addition, some illustrative examples are given at the end of this paper.


## 1. Introduction

Let $\mathbb{X}$ be a real Hilbert space and $A: \mathbb{X} \rightrightarrows \mathbb{X}, B: \mathbb{X} \rightrightarrows \mathbb{X}$ be two maximal monotone operators on $\mathbb{X}$. We consider the following principal problem: find an element $x$ in $\mathbb{X}$ such that

$$
\begin{equation*}
0 \in A(x)+B(x) . \tag{1.1}
\end{equation*}
$$

We denote the set of solutions of (1.1) by $(A+B)^{-1}(0)$.
The problem (1.1) was studied by various authors for example, see, $[1,2,3$, $4,7,8,9,10,11,12,14,15]$.

For all $\lambda>0$, we define the following function $f_{\lambda}: \mathbb{X}^{2} \rightarrow \mathbb{X}^{2}$ by

[^0]$$
f_{\lambda}(x, y)=\binom{J_{\lambda}^{A}(x)-\frac{x+y}{2}}{J_{\lambda}^{B}(y)-\frac{x+y}{2}}
$$
where $J_{\lambda}^{A}$ and $J_{\lambda}^{B}$ are respectively the resolvents of $A$ and $B$, which are defined as follows:
$$
J_{\lambda}^{A}(x):=(I+\lambda A)^{-1}(x) \quad \text { and } \quad J_{\lambda}^{B}(x):=(I+\lambda B)^{-1}(x)
$$
where $I$ is the identity operator in $\mathbb{X}$.
For all $\lambda>0$, we consider the problem:
\[

$$
\begin{equation*}
\min _{(x, y) \in \mathbb{X} \times \mathbb{X}}\left\|f_{\lambda}(x, y)\right\|=0 \tag{1.2}
\end{equation*}
$$

\]

It is well known that the second problem (1.2) has many algorithms to obtain one of its solutions(see [13]).

This paper is organized as follows: in the second section, we establish some notations and some properties of maximal monotone operator. In the third section, we will explain the relationship between the problems (1.1) and (1.2). In the last section, we use the Newton-Raphson algorithm to approximate a solution of (1.1).

## 2. Preliminaries

Let $\mathbb{X}$ be a real Hilbert space and $A: \mathbb{X} \rightrightarrows \mathbb{X}$ a set-valued map. We denote by $\operatorname{dom} A$ the domain of $A$ that is, $\operatorname{dom} A:=\{x \in \mathbb{X}: A x \neq \emptyset\}$, and the graph of $A$ is given by:

$$
\operatorname{graph} A:=\{(x, y) \in \mathbb{X} \times \mathbb{X}: x \in \operatorname{dom} A \text { and } y \in A x\}
$$

Definition 2.1. ([6]) The operator $A$ is said to be monotone, if:

$$
\left\langle y_{1}-y_{2}, x_{1}-x_{2}\right\rangle \geq 0 \quad \text { for all } \quad\left(x_{i}, y_{i}\right) \in \operatorname{graph} A, i=1,2
$$

Definition 2.2. ([6]) A monotone operator $A$ is called maximal, if its graph has not an extension to a graph of another monotone operator.

We can rephrase the definition of a maximal monotone operator in terms of its graph. If $A$ is a monotone operator, then $A$ is maximal if and only if the following conditions are equivalents:
(1) $u \in A(x)$,
(2) for all $(y, v) \in \operatorname{graph}(A), \quad\langle u-v, x-y\rangle \geq 0$.

Proposition 2.3. ([6]) Let $A: \mathbb{X} \rightrightarrows \mathbb{X}$ be a maximal monotone operator. Then for all $\lambda>0,(I+\lambda A)$ is surjective.

Proposition 2.4. ([6]) Let $A$ be a maximal monotone operator. For each $y \in \mathbb{X}$, there is a unique $x \in \operatorname{dom} A$ such that $y \in x+\lambda A(x)$.

For $\lambda>0$, the Yoshida approximation of a maximal monotone $A$ is given by

$$
A_{\lambda}(x):=\frac{1}{\lambda}\left(\mathrm{id}-J_{\lambda}^{A}\right)(x) .
$$

Proposition 2.5. ([5, 6]) For all element $y \in \mathbb{X}$, we have

$$
A_{\lambda}(y) \in A\left(J_{\lambda}(y)\right) .
$$

Definition 2.6. The sum $A+B$ is defined as:

$$
A(x)+B(x)=\{u+v / u \in A(x) \text { and } v \in B(x)\}
$$

Clearly $\operatorname{dom}(A+B)=\operatorname{dom} A \cap \operatorname{dom} B$.
Let $f: \mathbb{X} \rightarrow \mathbb{R} \cup\{\infty\}$ be a proper, convex and lower semicontinuous function. The subdifferential of $f, \partial f: \mathbb{X} \rightrightarrows \mathbb{X}$ is defined by

$$
\partial f(x)=\left\{x^{*} \in \mathbb{X}: f(y)-f(x) \geq\left\langle y-x, x^{*}\right\rangle, \quad \forall y \in \mathbb{X}\right\}
$$

It is known that $\partial f: \mathbb{X} \rightrightarrows \mathbb{X}$ is a maximal monotone operator on $\mathbb{X}$, and $0 \in \partial f\left(x^{*}\right)$ if and only if $x^{*}$ is a minimizer of $f$. Setting $\partial f=A$, it follows that solving the inclusion $0 \in A x$, in this case, is equivalent to solving for a minimizer of $f$.

## 3. Main results

Our goal in this section is to find the relation between the problems (1.1) and (1.2), we start with the following theorem.

Theorem 3.1. For all $\lambda>0$, the problem (1.2) has a solution if and only if (1.1) has a solution.

Proof. For all $\lambda>0$, if $(x, y)$ is a solution of (1.2), then

$$
\left\{\begin{array}{l}
J_{\lambda}^{A}(x)=\frac{x+y}{2}, \\
J_{\lambda}^{B}(y)=\frac{x+y}{2} .
\end{array}\right.
$$

Therefore,

$$
\left\{\begin{array}{l}
y=2 J_{\lambda}^{A}(x)-x, \\
2 J_{\lambda}^{B}(y)=x+y .
\end{array}\right.
$$

Replacing $y$ by $2 J_{\lambda}^{A}(x)-x$ in $2 J_{\lambda}^{B}(y)=x+y$, we get,

$$
J_{\lambda}^{B}\left(2 J_{\lambda}^{A}(x)-x\right)=J_{\lambda}^{A}(x) .
$$

So

$$
2 J_{\lambda}^{A}(x)-x \in J_{\lambda}^{A}(x)+\lambda B\left(J_{\lambda}^{A}(x)\right) .
$$

That means

$$
-A_{\lambda}(x) \in B\left(J_{\lambda}^{A}(x)\right)
$$

Since $A_{\lambda}(x) \in A\left(J_{\lambda}^{A}(x)\right)$, we obtain

$$
0 \in A\left(J_{\lambda}^{A}(x)\right)+B\left(J_{\lambda}^{A}(x)\right) .
$$

Consequently

$$
J_{\lambda}^{A}(x) \in(A+B)^{-1}(0) .
$$

Conversely, if $z \in(A+B)^{-1}(0)$, then $z \in \operatorname{dom} A$. This means that there exists an element $x^{\prime} \in \mathbb{X}$ such that $x^{\prime} \in A(z)$. So,

$$
x^{\prime}+z \in z+A(z) \text {, }
$$

and by the definition of the resolvent of $A$, we get

$$
z=J_{\lambda}^{A}\left(x^{\prime}+z\right) .
$$

We conclude that, for all $z \in(A+B)^{-1}(0)$, there exists $s=x^{\prime}+z \in \mathbb{X}$ such that $z=J_{\lambda}^{A}(s)$. Thus we have

$$
J_{\lambda}^{A}(s) \in(A+B)^{-1}(0) .
$$

Then

$$
0 \in A\left(J_{\lambda}^{A}(s)\right)+B\left(J_{\lambda}^{A}(s)\right) .
$$

Therefore, there exists $a \in \mathbb{X}$ such that

$$
\left\{\begin{array}{c}
a \in A\left(J_{\lambda}^{A}(s)\right), \\
-a \in B\left(J_{\lambda}^{A}(s)\right) .
\end{array}\right.
$$

Hence

$$
\left\{\begin{array}{c}
a+J_{\lambda}^{A}(s) \in J_{\lambda}^{A}(s)+A\left(J_{\lambda}^{A}(s)\right) \\
-a+J_{\lambda}^{A}(s) \in J_{\lambda}^{A}(s)+B\left(J_{\lambda}^{A}(s)\right) .
\end{array}\right.
$$

By the definition of the resolvent of $A$, we have

$$
\left\{\begin{array}{l}
J_{\lambda}^{A}(s)=J_{\lambda}^{A}\left(J_{\lambda}^{A}(s)+a\right), \\
J_{\lambda}^{A}(s)=J_{\lambda}^{B}\left(J_{\lambda}^{A}(s)-a\right) .
\end{array}\right.
$$

We can write $J_{\lambda}^{A}(s)$ as

$$
J_{\lambda}^{A}(s)=\frac{x+y}{2},
$$

where

$$
\left\{\begin{array}{l}
x=J_{\lambda}^{A}(s)+a, \\
y=J_{\lambda}^{A}(s)-a .
\end{array}\right.
$$

Then

$$
\left\{\begin{aligned}
J_{\lambda}^{A}(x) & =\frac{x+y}{2}, \\
J_{\lambda}^{B}(y) & =\frac{x+y}{2} .
\end{aligned}\right.
$$

Consequently, the problem (1.2) has a solution.

Remark 3.2. Let $\lambda>0, \mu>0, S_{\lambda}=\left\{(x, y) \in \mathbb{X} \times \mathbb{X}:\left\|f_{\lambda}(x, y)\right\|=0\right\}$ and $S_{\mu}=\left\{(x, y) \in \mathbb{X} \times \mathbb{X}:\left\|f_{\mu}(x, y)\right\|=0\right\}$. Then

$$
\left\{J_{\lambda}^{A}(x):(x, y) \in S_{\lambda}\right\}=\left\{J_{\mu}^{A}(x):(x, y) \in S_{\mu}\right\} .
$$

that is,

$$
(A+B)^{-1}(0)=\left\{J_{\lambda}^{A}(x):(x, y) \in S_{\lambda}\right\}, \quad \forall \lambda>0
$$

Theorem 3.3. For all $\lambda>0$, if $\theta_{\lambda}(a)=J_{\lambda}^{B}\left(2 J_{\lambda}^{A}(a)-a\right)-J_{\lambda}^{A}(a)+a$, then

$$
J_{\lambda}^{A}\left(S\left(\theta_{\lambda}\right)\right)=(A+B)^{-1}(0),
$$

where $S\left(\theta_{\lambda}\right)$ denotes the set of all fixed points of $\theta_{\lambda}$.
Proof. If $a$ is a fixed point for the function $\theta_{\lambda}$, we have

$$
\begin{aligned}
a=J_{\lambda}^{B}\left(2 J_{\lambda}^{A}(a)-a\right)-J_{\lambda}^{A}(a)+a & \Longleftrightarrow J_{\lambda}^{B}\left(2 J_{\lambda}^{A}(a)-a\right)=J_{\lambda}^{A}(a) \\
& \Longleftrightarrow 2 J_{\lambda}^{A}(a)-a \in J_{\lambda}^{A}(a)+\lambda B\left(J_{\lambda}^{A}(a)\right) \\
& \Longleftrightarrow \frac{1}{\lambda}\left(J_{\lambda}^{A}(a)-a\right) \in B\left(J_{\lambda}^{A}(a)\right) \\
& \Longleftrightarrow-A_{\lambda}(a) \in B\left(J_{\lambda}^{A}(a)\right) .
\end{aligned}
$$

Since $A_{\lambda}(a) \in A\left(J_{\lambda}^{A}(a)\right)$, we get $0 \in A\left(J_{\lambda}^{A}(a)\right)+B\left(J_{\lambda}^{A}(a)\right)$.
Example 3.4. Consider $A(x)=\{2 x\}, B(x)=\{4 x\}$ and $\mathbb{X}=\mathbb{R}$. We will find a solution of the following problem:

$$
\begin{equation*}
0 \in A(x)+B(x) \tag{3.1}
\end{equation*}
$$

Clearly, for all $\lambda>0$,

$$
J_{\lambda}^{A}(x)=\frac{x}{2 \lambda+1} \quad \text { and } \quad J_{\lambda}^{B}(x)=\frac{x}{4 \lambda+1} .
$$

We can define the function: $f_{\lambda}: \mathbb{R}^{2} \rightarrow \mathbb{R}^{2}$ by

$$
f_{\lambda}(x, y)=\binom{J_{\lambda}^{A}(x)-\frac{x+y}{2}}{J_{\lambda}^{B}(y)-\frac{x+y}{2}}=\binom{\frac{x}{2 \lambda+1}-\frac{x+y}{2}}{\frac{y}{4 \lambda+1}-\frac{x+y}{2}} .
$$

If $\left\|f_{\lambda}(x, y)\right\|=0$, then

$$
\left\{\begin{array}{l}
\frac{x}{2 \lambda+1}-\frac{x+y}{2}=0 \\
\frac{y}{4 \lambda+1}-\frac{x+y}{2}=0
\end{array}\right.
$$

Hence $(x, y)=(0,0)$. Apply Theorem (3.1), we obtain $J_{\lambda}^{A}(0)=0$ is a solution of (3.1).

## 4. Numerical example

4.1. Algorithm. We have:

$$
\begin{aligned}
f: & \mathbb{R}^{2} \rightarrow \mathbb{R}^{2} \\
& (x, y) \rightarrow f(x, y)=\binom{J_{\lambda}^{A}(x)-\frac{x+y}{2}}{J_{\lambda}^{B}(y)-\frac{x+y}{2}} .
\end{aligned}
$$

The Jacobian matrix of $f$ is :

$$
J_{f}(x, y)=\left(\begin{array}{cc}
\frac{\partial J_{\lambda}^{A}}{\partial x}(x)-\frac{1}{2} & -\frac{1}{2} \\
-\frac{1}{2} & \frac{\partial J_{\lambda}^{B}}{\partial y}(y)-\frac{1}{2}
\end{array}\right) .
$$

We calculate the determinant of $J_{f}(x, y)$ :

$$
\begin{aligned}
\operatorname{det} J_{f}(x, y) & =\left(\frac{\partial J_{\lambda}^{A}}{\partial x}(x)-\frac{1}{2}\right)\left(\frac{\partial J_{\lambda}^{B}}{\partial y}(y)-\frac{1}{2}\right)-\frac{1}{4} \\
& =\frac{\partial J_{\lambda}^{A}}{\partial x}(x) \frac{\partial J_{\lambda}^{B}}{\partial y}(y)-\frac{1}{2} \frac{\partial J_{\lambda}^{A}}{\partial x}(x)-\frac{1}{2} \frac{\partial J_{\lambda}^{B}}{\partial y}(y) .
\end{aligned}
$$

Newton-Raphson method, or Newton Method, is a powerful technique for solving the equations numerically:

$$
\left(x_{k+1}, y_{k+1}\right)=\left(x_{k}, y_{k}\right)-J_{f}^{-1}\left(x_{k}, y_{k}\right) f\left(x_{k}, y_{k}\right), \quad k=0,1,2, \ldots
$$

### 4.2. MATLAB programming.

```
clear all; close all; clc ; \(x=10\);
\(y=10\);
variable \(=[x ; y]\);
\(t=[0 ; 0]\);
delta \(=1\);
iteration \(=1\);
while (abs (delta) > \(1 e-20\) )
\(f=\left[J_{\lambda}^{A}(\right.\) variable \((1))+0.5 *(v a r i a b l e(1) * \operatorname{variable}(2)) ;\)
\(J_{\lambda}^{B}(\) variable \(\left.(1))+0.5 *(v a r i a b l e(1) * \operatorname{variable}(2))\right] ;\)
jacop \(=\left[\left(\frac{\partial J_{A}^{A}}{\partial x}(\right.\right.\) variable \(\left.(1))-0.5\right)(-0.5) ;\left((-0.5) \frac{\partial J_{B}^{B}}{\partial x}(\right.\) variable \(\left.\left.(1))-0.5\right)\right]\);
delta \(=(j a c o p) *(t-f)\);
variable \(=\) variable + delta ;
iteration \(=\) iteration +1 ;
end
format long
x_optimise \(=\) variable \((1)\)
\(y_{\text {_optimise }}=\) variable \((2)\)
```

Example 4.1. Let $f: \mathbb{R} \rightarrow \mathbb{R}$ and $g: \mathbb{R} \rightarrow \mathbb{R}$ be two convex functions defined as:

$$
f(x)=(2 x+1)^{2} \quad \text { and } \quad g(x)=(5 x+2)^{2} .
$$

Then

$$
\partial f(x)=\{8 x+4\} \quad \text { and } \quad \partial g(x)=\{50 x+20\} .
$$

It is known that $A=\partial f: \mathbb{R} \rightrightarrows \mathbb{R}$ and $B=\partial g: \mathbb{R} \rightrightarrows \mathbb{R}$ are maximal monotone operators on $\mathbb{R}$. Our problem,

$$
\begin{equation*}
0 \in \partial f(x)+\partial g(x) \tag{4.1}
\end{equation*}
$$

Then, for all $\lambda>0$,

$$
J_{\lambda}^{A}(x)=\frac{x-4 \lambda}{8 \lambda+1} \quad \text { and } \quad J_{\lambda}^{B}(x)=\frac{x-20 \lambda}{50 \lambda+1} .
$$

If we choose $\lambda=1$, we obtain by using the above program,

$$
(x, y)=(0.246,-1.073)
$$

and from Theorem 3.1, $J_{1}^{A}(0.246)=-0.417$ is an approximate solution of (4.1).

## References

[1] H.A. Abass, A.A. Mebawondu, O.K. Narain and J.K. Kim, Outer approximation method for zeros of sum of monotone operators and fixed point problems in Banach spaces, Nonlinear Funct. Anal. Appl., 26(3) (2021), 451-474.
[2] A. Adamu, J. Deepho, A.H. Ibrahim and A.B. Abubakar, Approximation of zeros of sum of monotone mappings with applications to variational inequality and image restoration problems, Nonlinear Funct. Anal. Appl., 26(2) (2021), 411-432.
[3] K. Afassinou, O.K. Narain and O.E. Otunuga Iterative algorithm for approximating solutions of Split Monotone Variational Inclusion, Variational inequality and fixed point problems in real Hilbert spaces, Nonlinear Funct. Anal. Appl., 25(3) (2020), 491-510.
[4] H. Attouch, J.B. Baillon and M. Théra, Variational Sum of Monotone Operators, J. Convex Anal., 1(1) (1994), 1-29.
[5] A. Beddani, An approximate solution of a differential inclusion with maximal monotone operator, J. Taibah Univ. Sci., 14(1) (2020), 1475-1481.
[6] H. Brezis, Opérateurs Maximaux Monotones et Semi-groupes de Contractions dans les Espaces de Hilbert, Math. Studies 5, North-Holland Ameri. Elsevier, 1973.
[7] C.E. Chidume, G.S. De Souza, O.M. Romanus and U.V. Nnyaba, A strong convergence theorem for maximal monotone operators in Banach spaces with applications, Carpathian J. Math., 36(2) (2020), 229-240.
[8] V. Dadashi and M. Postolache, Forward-backward splitting algorithm for fixed point problems and zeros of the sum of monotone operators, Arab. J. Math., 9 (2020), 89-99.
[9] A. Moudafi and M. Théra, Finding a Zero of The Sum of Two Maximal Monotone Operators, J. Opti.Theory Appl., 94(2) (1997), 425-448.
[10] F. Ogbuisi and I. Chinedu, Approximating a Zero of Sum of Two Monotone Operators Which Solves a Fixed Point Problem in Reflexive Banach Spaces, Numerical Funct. Anal. Opti., 41(146) (2020), 1-22.
[11] Y. Shehu, Convergence Results of Forward-Backward Algorithms for Sum of Monotone Operators in Banach Spaces, Results in Math., 74(138) (2019), 1-24.
[12] Y. Shehu, Q.L. Dong, L.L Liu and J.C. Yao, New strong convergence method for the sum of two maximal monotone operators, Opti. and Eng., 22 (2020), 2627-2653.
[13] J. Tang, J. Zhu, S.S. Chang, M. Liu and X. Li, A new modified proximal point algorithm for a finite family of minimization problem and fixed point for a finite family of demicontractive mappings in Hadamard spaces, Nonlinear Funct. Anal. Appl., 25(3) (2020), 563-577.
[14] M.M. Walaa and V. Lieven, Douglas-Rachford Splitting for the Sum of a Lipschitz Continuous and a Strongly Monotone Operator, J. Opti. Theory Appl., 183 (2019), 179-198.
[15] G.B. Wega and Z. Habtu, A strong convergence theorem for a zero of the sum of a finite family of maximally monotone mappings, Demonstratio Math., 53(1) (2020), 152-166.


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