

Convergence d'une méthode de résolution de systèmes d'inclusions monotones couplées

H. Attouch, Luis M. Briceño-Arias, P. L. Combettes

Laboratoire Jacques-Louis Lions
Équipe Combinatoire et Optimisation
Université Pierre et Marie Curie – Paris 6
75005 Paris, France

Porquerolles, 19 Octobre 2009

- 1 Notation
- 2 Introduction
- 3 Problem
- 4 Algorithm
- 5 Applications

- 1 Notation
- 2 Introduction
- 3 Problem
- 4 Algorithm
- 5 Applications

1 Notation

2 Introduction

3 Problem

4 Algorithm

5 Applications

Notation

Let \mathcal{H} be a Hilbert space. $\Gamma_0(\mathcal{H})$ denotes the class of proper lower semicontinuous convex functions from \mathcal{H} to $]-\infty, +\infty]$. For an operator $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$

- $\text{gra}(A) = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid u \in Ax\}$ is the graph of A .
- A is called **monotone** if, for every (x, u) and (y, v) in $\text{gra}(A)$, we have

$$\langle x - y \mid u - v \rangle \geq 0.$$

In addition, A is **maximal** if its graph is maximal (in the sense of inclusions) between the monotone operators in \mathcal{H} .

- The resolvent operator of A , $J_A = (\text{Id} + A)^{-1}$, is single-valued.

Notation

Let \mathcal{H} be a Hilbert space. $\Gamma_0(\mathcal{H})$ denotes the class of proper lower semicontinuous convex functions from \mathcal{H} to $] -\infty, +\infty]$. For an operator $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$

- $\text{gra}(A) = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid u \in Ax\}$ is the graph of A .
- A is called **monotone** if, for every (x, u) and (y, v) in $\text{gra}(A)$, we have

$$\langle x - y \mid u - v \rangle \geq 0.$$

In addition, A is **maximal** if its graph is maximal (in the sense of inclusions) between the monotone operators in \mathcal{H} .

- The resolvent operator of A , $J_A = (\text{Id} + A)^{-1}$, is single-valued.

Notation

Let \mathcal{H} be a Hilbert space. $\Gamma_0(\mathcal{H})$ denotes the class of proper lower semicontinuous convex functions from \mathcal{H} to $] -\infty, +\infty]$. For an operator $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$

- $\text{gra}(A) = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid u \in Ax\}$ is the graph of A .
- A is called **monotone** if, for every (x, u) and (y, v) in $\text{gra}(A)$, we have

$$\langle x - y \mid u - v \rangle \geq 0.$$

In addition, A is **maximal** if its graph is maximal (in the sense of inclusions) between the monotone operators in \mathcal{H} .

- The resolvent operator of A , $J_A = (\text{Id} + A)^{-1}$, is single-valued.

1 Notation

2 Introduction

3 Problem

4 Algorithm

5 Applications

Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

Best approximation problem

$$\min_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$

is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$

Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

Best approximation problem

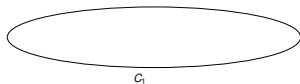
$$\min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} \iota_{C_1}(x_1) + \iota_{C_2}(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$



is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$



Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

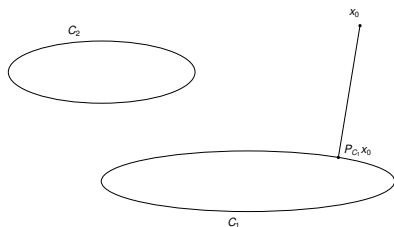
Best approximation problem

$$\min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} \iota_{C_1}(x_1) + \iota_{C_2}(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$

is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$



Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

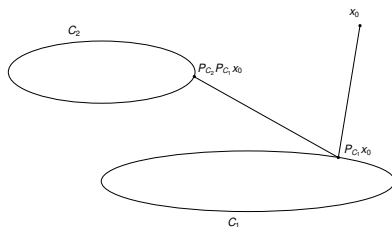
Best approximation problem

$$\min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} \iota_{C_1}(x_1) + \iota_{C_2}(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$

is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$



Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

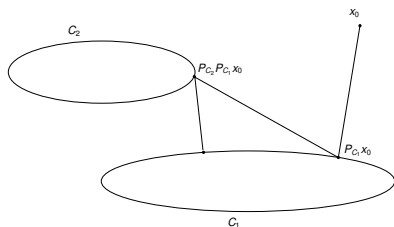
Best approximation problem

$$\min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} \iota_{C_1}(x_1) + \iota_{C_2}(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$

is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$



Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

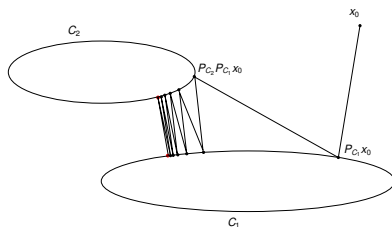
Best approximation problem

$$\min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} \iota_{C_1}(x_1) + \iota_{C_2}(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$

is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$



Motivation

Let C_1 y C_2 be nonempty closed convex subsets of a real Hilbert space \mathcal{H} .

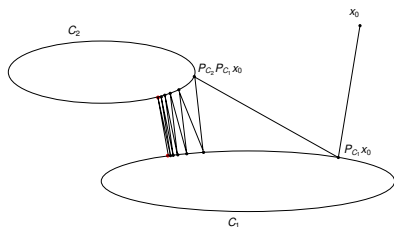
Best approximation problem

$$\min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} \iota_{C_1}(x_1) + \iota_{C_2}(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in N_{C_1} x_1 + x_1 - x_2 \\ 0 \in N_{C_2} x_2 - x_1 + x_2 \end{cases}$$

is equivalent to

$$\begin{cases} x_1 = P_{C_1} x_2 \\ x_2 = P_{C_2} x_1 \end{cases}$$



Motivation

Let f_1 y f_2 in $\Gamma_0(\mathcal{H})$.

Acker and Prestel'80

$$(\mathcal{P}_1) \quad \min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + x_1 - x_2 \\ 0 \in \partial f_2(x_2) - x_1 + x_2 \end{cases} \Leftrightarrow \begin{cases} x_1 = (\text{Id} + \partial f_1)^{-1} x_2 \\ x_2 = (\text{Id} + \partial f_2)^{-1} x_1. \end{cases}$$

Let $x_{2,0} \in \mathcal{H}$ and set

$$\begin{cases} x_{1,n+1} = (\text{Id} + \partial f_1)^{-1} x_{2,n} \\ x_{2,n+1} = (\text{Id} + \partial f_2)^{-1} x_{1,n+1} \end{cases} \Rightarrow \begin{cases} x_{1,n} \rightarrow x_1, x_{2,n} \rightarrow x_2 \\ (x_1, x_2) \text{ is solution of } (\mathcal{P}_1). \end{cases}$$

Motivation

Let f_1 y f_2 in $\Gamma_0(\mathcal{H})$.

Acker and Prestel'80

$$(\mathcal{P}_1) \quad \min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + x_1 - x_2 \\ 0 \in \partial f_2(x_2) - x_1 + x_2 \end{cases} \Leftrightarrow \begin{cases} x_1 = (\text{Id} + \partial f_1)^{-1} x_2 \\ x_2 = (\text{Id} + \partial f_2)^{-1} x_1. \end{cases}$$

Let $x_{2,0} \in \mathcal{H}$ and set

$$\begin{cases} x_{1,n+1} = (\text{Id} + \partial f_1)^{-1} x_{2,n} \\ x_{2,n+1} = (\text{Id} + \partial f_2)^{-1} x_{1,n+1} \end{cases} \Rightarrow \begin{cases} x_{1,n} \rightarrow x_1, x_{2,n} \rightarrow x_2 \\ (x_1, x_2) \text{ is solution of } (\mathcal{P}_1). \end{cases}$$

Motivation

Let f_1 y f_2 in $\Gamma_0(\mathcal{H})$.

Acker and Prestel'80

$$(\mathcal{P}_1) \quad \min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + x_1 - x_2 \\ 0 \in \partial f_2(x_2) - x_1 + x_2 \end{cases} \Leftrightarrow \begin{cases} x_1 = (\text{Id} + \partial f_1)^{-1} x_2 \\ x_2 = (\text{Id} + \partial f_2)^{-1} x_1. \end{cases}$$

Let $x_{2,0} \in \mathcal{H}$ and set

$$\begin{cases} x_{1,n+1} = (\text{Id} + \partial f_1)^{-1} x_{2,n} \\ x_{2,n+1} = (\text{Id} + \partial f_2)^{-1} x_{1,n+1} \end{cases} \Rightarrow \begin{cases} x_{1,n} \rightarrow x_1, x_{2,n} \rightarrow x_2 \\ (x_1, x_2) \text{ is solution of } (\mathcal{P}_1). \end{cases}$$

Motivation

Let f_1 y f_2 in $\Gamma_0(\mathcal{H})$.

Acker and Prestel'80

$$(\mathcal{P}_1) \quad \min_{x_1 \in \mathcal{H}, x_2 \in \mathcal{H}} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|x_1 - x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + x_1 - x_2 \\ 0 \in \partial f_2(x_2) - x_1 + x_2 \end{cases} \Leftrightarrow \begin{cases} x_1 = (\text{Id} + \partial f_1)^{-1} x_2 \\ x_2 = (\text{Id} + \partial f_2)^{-1} x_1. \end{cases}$$

Let $x_{2,0} \in \mathcal{H}$ and set

$$\begin{cases} x_{1,n+1} = (\text{Id} + \partial f_1)^{-1} x_{2,n} \\ x_{2,n+1} = (\text{Id} + \partial f_2)^{-1} x_{1,n+1} \end{cases} \Rightarrow \begin{cases} x_{1,n} \rightarrow x_1, x_{2,n} \rightarrow x_2 \\ (x_1, x_2) \text{ is solution of } (\mathcal{P}_1). \end{cases}$$

Motivation

- Let A_1, A_2 be maximal monotone operators defined on a Hilbert space \mathcal{H} .

Monotone inclusion

$$(\mathcal{P}_2) \quad \begin{cases} 0 \in A_1 x_1 + x_1 - x_2 \\ 0 \in A_2 x_2 - x_1 + x_2 \end{cases}$$

- Let $x_{1,0} \in \mathcal{H}$. For all $n \in \mathbb{N}$

$$\begin{cases} x_{2,n} = (\text{Id} + A_2)^{-1} x_{1,n} \\ x_{1,n+1} = (\text{Id} + A_1)^{-1} x_{2,n} \end{cases}$$
- $x_{1,n} \rightarrow x_1$ and $x_{2,n} \rightarrow x_2$ where (x_1, x_2) is solution of (\mathcal{P}_2) (Bauschke et al.'05).

Motivation

- Let A_1, A_2 be maximal monotone operators defined on a Hilbert space \mathcal{H} .

Monotone inclusion

$$(\mathcal{P}_2) \quad \begin{cases} 0 \in A_1 x_1 + x_1 - x_2 \\ 0 \in A_2 x_2 - x_1 + x_2 \end{cases}$$

- Let $x_{1,0} \in \mathcal{H}$. For all $n \in \mathbb{N}$

$$\begin{cases} x_{2,n} = (\text{Id} + A_2)^{-1} x_{1,n} \\ x_{1,n+1} = (\text{Id} + A_1)^{-1} x_{2,n} \end{cases}$$
- $x_{1,n} \rightharpoonup x_1$ and $x_{2,n} \rightharpoonup x_2$ where (x_1, x_2) is solution of (\mathcal{P}_2) (Bauschke et al.'05).

Motivation

- Let A_1, A_2 be maximal monotone operators defined on a Hilbert space \mathcal{H} .

Monotone inclusion

$$(\mathcal{P}_2) \quad \begin{cases} 0 \in A_1 x_1 + x_1 - x_2 \\ 0 \in A_2 x_2 - x_1 + x_2 \end{cases}$$

- Let $x_{1,0} \in \mathcal{H}$. For all $n \in \mathbb{N}$

$$\begin{cases} x_{2,n} = (\text{Id} + A_2)^{-1} x_{1,n} \\ x_{1,n+1} = (\text{Id} + A_1)^{-1} x_{2,n} \end{cases}$$
- $x_{1,n} \rightharpoonup x_1$ and $x_{2,n} \rightharpoonup x_2$ where (x_1, x_2) is solution of (\mathcal{P}_2) (Bauschke et al.'05).

Motivation

Let \mathcal{H}_1 , \mathcal{H}_2 and \mathcal{G} be real Hilbert spaces, and, for every $i \in \{1, 2\}$, let $f_i \in \Gamma_0(\mathcal{H}_i)$ and let $L_i: \mathcal{H}_i \rightarrow \mathcal{G}$ be linear and bounded.

Attouch et al.'08

$$(\mathcal{P}_3) \quad \min_{x_1 \in \mathcal{H}_1, x_2 \in \mathcal{H}_2} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|L_1 x_1 - L_2 x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + L_1^* L_1 x_1 - L_1^* L_2 x_2 \\ 0 \in \partial f_2(x_2) - L_2^* L_1 x_1 + L_2^* L_2 x_2 \end{cases}$$

An alternating algorithm is proposed, which generates sequences $(x_{1,n})_{n \in \mathbb{N}}$ and $(x_{2,n})_{n \in \mathbb{N}}$ that weakly converges to a solution of (\mathcal{P}_3) .

Motivation

Let \mathcal{H}_1 , \mathcal{H}_2 and \mathcal{G} be real Hilbert spaces, and, for every $i \in \{1, 2\}$, let $f_i \in \Gamma_0(\mathcal{H}_i)$ and let $L_i: \mathcal{H}_i \rightarrow \mathcal{G}$ be linear and bounded.

Attouch et al.'08

$$(\mathcal{P}_3) \quad \min_{x_1 \in \mathcal{H}_1, x_2 \in \mathcal{H}_2} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|L_1 x_1 - L_2 x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + L_1^* L_1 x_1 - L_1^* L_2 x_2 \\ 0 \in \partial f_2(x_2) - L_2^* L_1 x_1 + L_2^* L_2 x_2 \end{cases}$$

An alternating algorithm is proposed, which generates sequences $(x_{1,n})_{n \in \mathbb{N}}$ and $(x_{2,n})_{n \in \mathbb{N}}$ that weakly converges to a solution of (\mathcal{P}_3) .

Motivation

Let \mathcal{H}_1 , \mathcal{H}_2 and \mathcal{G} be real Hilbert spaces, and, for every $i \in \{1, 2\}$, let $f_i \in \Gamma_0(\mathcal{H}_i)$ and let $L_i: \mathcal{H}_i \rightarrow \mathcal{G}$ be linear and bounded.

Attouch et al.'08

$$(\mathcal{P}_3) \quad \min_{x_1 \in \mathcal{H}_1, x_2 \in \mathcal{H}_2} f_1(x_1) + f_2(x_2) + \frac{1}{2} \|L_1 x_1 - L_2 x_2\|^2$$

$$\begin{cases} 0 \in \partial f_1(x_1) + L_1^* L_1 x_1 - L_1^* L_2 x_2 \\ 0 \in \partial f_2(x_2) - L_2^* L_1 x_1 + L_2^* L_2 x_2 \end{cases}$$

An alternating algorithm is proposed, which generates sequences $(x_{1,n})_{n \in \mathbb{N}}$ and $(x_{2,n})_{n \in \mathbb{N}}$ that weakly converges to a solution of (\mathcal{P}_3) .

Our goal is...

- Extension to m variables
- More general couplings
- Different Hilbert spaces

Our goal is...

- Extension to m variables
- More general couplings
- Different Hilbert spaces

Our goal is...

- Extension to m variables
- More general couplings
- Different Hilbert spaces

- 1 Notation
- 2 Introduction
- 3 Problem**
- 4 Algorithm
- 5 Applications

General Problem

- For every $i \in \{1, \dots, m\}$:
 - \mathcal{H}_i is a real Hilbert space and denote $\mathcal{H} = \mathcal{H}_1 \oplus \dots \oplus \mathcal{H}_m$.
 - A_i is a maximal monotone operator from \mathcal{H}_i to $2^{\mathcal{H}_i}$
 - B_i is an operator from \mathcal{H} to \mathcal{H}_i (coupling) such that $(B_i)_{1 \leq i \leq m}$ satisfy

$$\begin{aligned}
 & (\forall (x_1, \dots, x_m) \in \mathcal{H}_1 \times \dots \times \mathcal{H}_m) (\forall (y_1, \dots, y_m) \in \mathcal{H}_1 \times \dots \times \mathcal{H}_m) \\
 & \quad \sum_{i=1}^m \langle B_i(x_1, \dots, x_m) - B_i(y_1, \dots, y_m) \mid x_i - y_i \rangle \\
 & \quad \geq \beta \sum_{i=1}^m \|B_i(x_1, \dots, x_m) - B_i(y_1, \dots, y_m)\|^2 \quad (C)
 \end{aligned}$$

for some $\beta \in]0, +\infty[$

General Problem

Un-coupled equilibrium problem

Find $(x_j)_{1 \leq j \leq m} \in \mathcal{H}$ such that

$$\left\{ \begin{array}{l} 0 \in A_1 x_1 \\ 0 \in A_2 x_2 \\ \vdots \\ 0 \in A_m x_m \end{array} \right.$$

General Problem

Coupled equilibrium problem

Find $(x_i)_{1 \leq i \leq m} \in \mathcal{H}$ such that

$$(P) \begin{cases} 0 \in A_1 x_1 + B_1(x_1, \dots, x_m) \\ 0 \in A_2 x_2 + B_2(x_1, \dots, x_m) \\ \vdots \\ 0 \in A_m x_m + B_m(x_1, \dots, x_m) \end{cases}$$

- 1 Notation
- 2 Introduction
- 3 Problem
- 4 Algorithm**
- 5 Applications

Algorithm and weakly convergence

- For every $n \in \mathbb{N}$ and every $i \in \{1, \dots, m\}$, set

Algorithm

$$x_{i,n+1} = \lambda_{i,n}x_{i,n} + (1 - \lambda_{i,n}) \left(J_{\gamma_n A_i} (x_{i,n} - \gamma_n B_i(x_{1,n}, \dots, x_{m,n})) \right)$$

- where for every $i \in \{1, \dots, m\}$
 - $(\gamma_n)_{n \in \mathbb{N}}$ be a sequence in $]0, 2\beta[$
 - $(\lambda_n)_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$.
 - $x_{i,0} \in \mathcal{H}_i$
 - $(\lambda_{i,n})_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$ such that $\sum_{n \in \mathbb{N}} |\lambda_{i,n} - \lambda_n| < +\infty$.
 - Proximity conditions $J_{A_i, \gamma} \leftrightarrow J_{A_i}$ and $B_{i, \gamma} \leftrightarrow B_i$
- The algorithm gives a sequence $((x_{i,n})_{1 \leq i \leq m})_{n \in \mathbb{N}}$ that converges weakly to a solution $(x_i)_{1 \leq i \leq m}$ to (P)

Algorithm and weakly convergence

- For every $n \in \mathbb{N}$ and every $i \in \{1, \dots, m\}$, set

Algorithm

$$x_{i,n+1} = \lambda_{i,n}x_{i,n} + (1 - \lambda_{i,n}) \left(J_{\gamma_n A_i} (x_{i,n} - \gamma_n B_i(x_{1,n}, \dots, x_{m,n})) \right)$$

- where for every $i \in \{1, \dots, m\}$
 - $(\gamma_n)_{n \in \mathbb{N}}$ be a sequence in $]0, 2\beta[$
 - $(\lambda_n)_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$.
 - $x_{i,0} \in \mathcal{H}_i$
 - $(\lambda_{i,n})_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$ such that $\sum_{n \in \mathbb{N}} |\lambda_{i,n} - \lambda_n| < +\infty$.
 - Proximity conditions $J_{A_{i,n}} \leftrightarrow J_{A_i}$ and $B_{i,n} \leftrightarrow B_i$
- The algorithm gives a sequence $((x_{i,n})_{1 \leq i \leq m})_{n \in \mathbb{N}}$ that converges weakly to a solution $(x_i)_{1 \leq i \leq m}$ to (P)

Algorithm and weakly convergence

- For every $n \in \mathbb{N}$ and every $i \in \{1, \dots, m\}$, set

Algorithm

$$x_{i,n+1} = \lambda_{i,n}x_{i,n} + (1 - \lambda_{i,n}) \left(J_{\gamma_n A_{i,n}} \left(x_{i,n} - \gamma_n B_{i,n}(x_{1,n}, \dots, x_{m,n}) \right) \right)$$

- where for every $i \in \{1, \dots, m\}$
 - $(\gamma_n)_{n \in \mathbb{N}}$ be a sequence in $]0, 2\beta[$
 - $(\lambda_n)_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$.
 - $x_{i,0} \in \mathcal{H}_i$
 - $(\lambda_{i,n})_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$ such that $\sum_{n \in \mathbb{N}} |\lambda_{i,n} - \lambda_n| < +\infty$.
 - Proximity conditions $J_{A_{i,n}} \leftrightarrow J_{A_i}$ and $B_{i,n} \leftrightarrow B_i$
- The algorithm gives a sequence $((x_{i,n})_{1 \leq i \leq m})_{n \in \mathbb{N}}$ that converges weakly to a solution $(x_i)_{1 \leq i \leq m}$ to (P)

Algorithm and weakly convergence

- For every $n \in \mathbb{N}$ and every $i \in \{1, \dots, m\}$, set

Algorithm

$$x_{i,n+1} = \lambda_{i,n}x_{i,n} + (1 - \lambda_{i,n}) \left(J_{\gamma_n A_{i,n}} \left(x_{i,n} - \gamma_n B_{i,n}(x_{1,n}, \dots, x_{m,n}) \right) \right)$$

- where for every $i \in \{1, \dots, m\}$
 - $(\gamma_n)_{n \in \mathbb{N}}$ be a sequence in $]0, 2\beta[$
 - $(\lambda_n)_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$.
 - $x_{i,0} \in \mathcal{H}_i$
 - $(\lambda_{i,n})_{n \in \mathbb{N}}$ be a sequence in $[0, 1[$ such that $\sum_{n \in \mathbb{N}} |\lambda_{i,n} - \lambda_n| < +\infty$.
 - Proximity conditions $J_{A_{i,n}} \leftrightarrow J_{A_i}$ and $B_{i,n} \leftrightarrow B_i$
- The algorithm gives a sequence $((x_{i,n})_{1 \leq i \leq m})_{n \in \mathbb{N}}$ that converges weakly to a solution $(x_i)_{1 \leq i \leq m}$ to (P)

Strong convergence

Suppose that, for some $i \in \{1, \dots, m\}$, one of the following holds

- \mathcal{H}_i is finite dimensional
- A_i is uniformly monotone, i.e. there exists a nondecreasing function $\phi: [0, +\infty[\rightarrow [0, +\infty]$, which vanishes only at 0, such that

$$(\forall (x, u) \in \text{gra}A_i)(\forall (y, v) \in \text{gra}A_i) \\ \langle x - y \mid u - v \rangle \geq \phi(\|x - y\|)$$

- J_{A_i} is compact
- $\text{dom}A_i$ is boundedly relatively compact

Then $x_{j,n} \rightarrow x_j$.

Strong convergence

Suppose that, for some $i \in \{1, \dots, m\}$, one of the following holds

- \mathcal{H}_i is finite dimensional
- A_i is uniformly monotone, i.e. there exists a nondecreasing function $\phi: [0, +\infty[\rightarrow [0, +\infty]$, which vanishes only at 0, such that

$$(\forall (x, u) \in \text{gra}A_i)(\forall (y, v) \in \text{gra}A_i) \\ \langle x - y \mid u - v \rangle \geq \phi(\|x - y\|)$$

- J_{A_i} is compact
- $\text{dom}A_i$ is boundedly relatively compact

Then $x_{j,n} \rightarrow x_j$.

Strong convergence

Suppose that, for some $i \in \{1, \dots, m\}$, one of the following holds

- \mathcal{H}_i is finite dimensional
- A_i is uniformly monotone, i.e. there exists a nondecreasing function $\phi: [0, +\infty[\rightarrow [0, +\infty]$, which vanishes only at 0, such that

$$(\forall (x, u) \in \text{gra}A_i)(\forall (y, v) \in \text{gra}A_i) \\ \langle x - y \mid u - v \rangle \geq \phi(\|x - y\|)$$

- J_{A_i} is compact
- $\text{dom}A_i$ is boundedly relatively compact

Then $x_{i,n} \rightarrow x_i$.

Strong convergence

Suppose that, one of the following holds

- There exists $\phi: [0, +\infty[\rightarrow [0, +\infty]$ nondecreasing that vanishes only at 0 such that, for every $(x_i)_{1 \leq i \leq m}$ and $(y_i)_{1 \leq i \leq m}$ in $\mathcal{H}_1 \times \cdots \times \mathcal{H}_m$

$$\sum_{i=1}^m \langle B_i(x_1, \dots, x_m) - B_i(y_1, \dots, y_m) \mid y_i - x_i \rangle \geq \phi \left(\sum_{i=1}^m \|x_i - y_i\|^2 \right)$$

- The set of solutions to Problem (P) has a nonempty interior

Then, for every $i \in \{1, \dots, m\}$, $x_{i,n} \rightarrow x_i$.

Strong convergence

Suppose that, one of the following holds

- There exists $\phi: [0, +\infty[\rightarrow [0, +\infty]$ nondecreasing that vanishes only at 0 such that, for every $(x_i)_{1 \leq i \leq m}$ and $(y_i)_{1 \leq i \leq m}$ in $\mathcal{H}_1 \times \cdots \times \mathcal{H}_m$

$$\begin{aligned} \sum_{i=1}^m \langle B_i(x_1, \dots, x_m) - B_i(y_1, \dots, y_m) \mid y_i - x_i \rangle \\ \geq \phi \left(\sum_{i=1}^m \|x_i - y_i\|^2 \right) \end{aligned}$$

- The set of solutions to Problem (P) has a nonempty interior

Then, for every $i \in \{1, \dots, m\}$, $x_{i,n} \rightarrow x_i$.

Strong convergence

Suppose that, one of the following holds

- There exists $\phi: [0, +\infty[\rightarrow [0, +\infty]$ nondecreasing that vanishes only at 0 such that, for every $(x_i)_{1 \leq i \leq m}$ and $(y_i)_{1 \leq i \leq m}$ in $\mathcal{H}_1 \times \cdots \times \mathcal{H}_m$

$$\begin{aligned} \sum_{i=1}^m \langle B_i(x_1, \dots, x_m) - B_i(y_1, \dots, y_m) \mid y_i - x_i \rangle \\ \geq \phi \left(\sum_{i=1}^m \|x_i - y_i\|^2 \right) \end{aligned}$$

- The set of solutions to Problem (P) has a nonempty interior

Then, for every $i \in \{1, \dots, m\}$, $x_{i,n} \rightarrow x_i$.

- 1 Notation
- 2 Introduction
- 3 Problem
- 4 Algorithm
- 5 Applications**

Coupling evolution equations

Let $T \in]0, +\infty[$ and, for every $i \in \{1, \dots, m\}$, let H_i be a Hilbert space.

The problem is to find functions $(x_i)_{1 \leq i \leq m}$ in $W^{1,2}([0, T], H_1) \times \dots \times W^{1,2}([0, T], H_m)$ such that, a.e. in $t \in]0, T[$

Coupling evolution problem (CEP)

$$(\forall i \in \{1, \dots, m\}) \quad \begin{cases} -x_i'(t) \in \partial f_i(x_i(t)) + D_i(x_1(t), \dots, x_m(t)); \\ x_i(0) = x_i(T), \end{cases}$$

where, for every $i \in \{1, \dots, m\}$,

- $f_i \in \Gamma_0(H_i)$
- $D_i: H_1 \times \dots \times H_m \rightarrow H_i$ is such that $(D_i)_{1 \leq i \leq m}$ satisfy (C).

Coupling evolution equations

Let $T \in]0, +\infty[$ and, for every $i \in \{1, \dots, m\}$, let H_i be a Hilbert space.

The problem is to find functions $(x_i)_{1 \leq i \leq m}$ in $W^{1,2}([0, T], H_1) \times \dots \times W^{1,2}([0, T], H_m)$ such that, a.e. in $t \in]0, T[$

Coupling evolution problem (CEP)

$$(\forall i \in \{1, \dots, m\}) \quad \begin{cases} -x_i'(t) \in \partial f_i(x_i(t)) + D_i(x_1(t), \dots, x_m(t)); \\ x_i(0) = x_i(T), \end{cases}$$

where, for every $i \in \{1, \dots, m\}$,

- $f_i \in \Gamma_0(H_i)$
- $D_i: H_1 \times \dots \times H_m \rightarrow H_i$ is such that $(D_i)_{1 \leq i \leq m}$ satisfy (C).

Coupling evolution equations

For every $i \in \{1, \dots, m\}$, let

- $\mathcal{H}_i = L^2([0, T]; H_i)$,
- $\mathcal{W}_i = \{x \in \mathcal{C}([0, T]; H_i) \cap W^{1,2}([0, T]; H_i) \mid x(T) = x(0)\}$,
- $A_i: \mathcal{H}_i \rightarrow 2^{\mathcal{H}_i}$,

$$A_i x = \begin{cases} \{u \in \mathcal{H}_i \mid u(t) \in x'(t) + \partial f_i(x(t)) \text{ a.e. in }]0, T[\}, & \text{if } x \in \mathcal{W}_i; \\ \emptyset, & \text{otherwise.} \end{cases}$$

is maximal monotone,

- $B_i: \mathcal{H}_1 \times \dots \times \mathcal{H}_m \rightarrow \mathcal{H}_i$,

$$\begin{aligned} B_i(x_1, \dots, x_m): [0, T] &\rightarrow H_i \\ t &\mapsto D_i(x_1(t), \dots, x_m(t)) \end{aligned}$$

is such that $(B_i)_{1 \leq i \leq m}$ satisfy (C) as well.

Coupling evolution equations

For every $i \in \{1, \dots, m\}$, let

- $\mathcal{H}_i = L^2([0, T]; H_i)$,
- $\mathcal{W}_i = \{x \in \mathcal{C}([0, T]; H_i) \cap W^{1,2}([0, T]; H_i) \mid x(T) = x(0)\}$,
- $A_i: \mathcal{H}_i \rightarrow 2^{\mathcal{H}_i}$,

$$A_i x = \begin{cases} \{u \in \mathcal{H}_i \mid u(t) \in x'(t) + \partial f_i(x(t)) \text{ a.e. in }]0, T[\}, & \text{if } x \in \mathcal{W}_i; \\ \emptyset, & \text{otherwise.} \end{cases}$$

is maximal monotone,

- $B_i: \mathcal{H}_1 \times \dots \times \mathcal{H}_m \rightarrow \mathcal{H}_i$,

$$B_i(x_1, \dots, x_m): [0, T] \rightarrow H_i$$

$$t \mapsto D_i(x_1(t), \dots, x_m(t))$$

is such that $(B_i)_{1 \leq i \leq m}$ satisfy (C) as well.

Coupling evolution equations

For every $i \in \{1, \dots, m\}$, let

- $\mathcal{H}_i = L^2([0, T]; H_i)$,
- $\mathcal{W}_i = \{x \in \mathcal{C}([0, T]; H_i) \cap W^{1,2}([0, T]; H_i) \mid x(T) = x(0)\}$,
- $A_i: \mathcal{H}_i \rightarrow 2^{\mathcal{H}_i}$,

$$A_i x = \begin{cases} \{u \in \mathcal{H}_i \mid u(t) \in x'(t) + \partial f_i(x(t)) \text{ a.e. in }]0, T[\}, & \text{if } x \in \mathcal{W}_i; \\ \emptyset, & \text{otherwise.} \end{cases}$$

is maximal monotone,

- $B_i: \mathcal{H}_1 \times \dots \times \mathcal{H}_m \rightarrow \mathcal{H}_i$,

$$\begin{aligned} B_i(x_1, \dots, x_m): [0, T] &\rightarrow H_i \\ t &\mapsto D_i(x_1(t), \dots, x_m(t)) \end{aligned}$$

is such that $(B_i)_{1 \leq i \leq m}$ satisfy (C) as well.

Coupling evolution equations

Problem (*CEP*) is equivalent to (*P*) and Algorithm becomes (we set errors to 0 and $\lambda_n \equiv 0$), for every $i \in \{1, \dots, m\}$ and a.e. in $t \in]0, T[$,

Algorithm (*CEP*)

$$(A) \quad x_{i,n}(t)/\gamma_n - D_i(x_{1,n}(t), \dots, x_{m,n}(t)) \\ \in x'_{i,n+1}(t) + \partial f_i(x_{i,n+1}(t)) + x_{i,n+1}(t)/\gamma_n,$$

where

- $(\gamma_n)_{n \in \mathbb{N}}$ in $]0, 2\beta[$
- $(\forall i \in \{1, \dots, m\}) \quad x_{i,0} \in W^{1,2}([0, T]; H_i)$.

Coupling evolution equations

Problem (CEP) is equivalent to (P) and Algorithm becomes (we set errors to 0 and $\lambda_n \equiv 0$), for every $i \in \{1, \dots, m\}$ and a.e. in $t \in]0, T[$,

Algorithm (CEP)

$$(A) \quad x_{i,n}(t)/\gamma_n - D_i(x_{1,n}(t), \dots, x_{m,n}(t)) \\ \in x'_{i,n+1}(t) + \partial(f_i + \frac{1}{2\gamma_n} \|\cdot\|^2)(x_{i,n+1}(t)),$$

where

- $(\gamma_n)_{n \in \mathbb{N}}$ in $]0, 2\beta[$
- $(\forall i \in \{1, \dots, m\}) \quad x_{i,0} \in W^{1,2}([0, T]; H_i)$.

There exists an unique solution of the inclusion (A) (Brezis'73).

The algorithm converges weakly to the solution of (CEP).

Coupling evolution equations

Problem (CEP) is equivalent to (P) and Algorithm becomes (we set errors to 0 and $\lambda_n \equiv 0$), for every $i \in \{1, \dots, m\}$ and a.e. in $t \in]0, T[$,

Algorithm (CEP)

$$(A) \quad x_{i,n}(t)/\gamma_n - D_i(x_{1,n}(t), \dots, x_{m,n}(t)) \\ \in x'_{i,n+1}(t) + \partial(f_i + \frac{1}{2\gamma_n} \|\cdot\|^2)(x_{i,n+1}(t)),$$

where

- $(\gamma_n)_{n \in \mathbb{N}}$ in $]0, 2\beta[$
- $(\forall i \in \{1, \dots, m\}) \quad x_{i,0} \in W^{1,2}([0, T]; H_i)$.

There exists an unique solution of the inclusion (A) (Brezis'73).

The algorithm converges weakly to the solution of (CEP).

Network Flows

- Directed graph (transport network).
- m types of flows, M links, N paths, Q origin-destination pairs. L is a $M \times N$ binary matrix (L_{jl} is 1 if link j belongs to path l and 0 otherwise).
- For every $i \in \{1, \dots, m\}$, $k \in \{1, \dots, Q\}$
 - $x_i \in \mathbb{R}^N$: flow type i on all the routes.
 - d_{ik} : total flow type i that transit from the origin to the destination of k .
 - N_k : number of paths linking k .
 - $C_i = \{(\eta_l)_{1 \leq l \leq N} \in [0, +\infty[^N \mid \forall k \sum_{l \in N_k} \eta_l = d_{ik}\}$
- For every $j \in \{1, \dots, M\}$,
 - $\phi_j: \mathbb{R} \rightarrow [0, +\infty[$: strictly increasing τ -Lipschitz continuous function (cost of transiting on link j).
 - $h_j: \mathbb{R}^{Nm} \rightarrow \mathbb{R}: (x_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m (Lx_i)^\top e_j$ total flow through link j .

Network Flows

- Directed graph (transport network).
- m types of flows, M links, N paths, Q origin-destination pairs. L is a $M \times N$ binary matrix (L_{jl} is 1 if link j belongs to path l and 0 otherwise).
- For every $i \in \{1, \dots, m\}$, $k \in \{1, \dots, Q\}$
 - $x_i \in \mathbb{R}^N$: flow type i on all the routes.
 - d_{ik} : total flow type i that transit from the origin to the destination of k .
 - N_k : number of paths linking k .
 - $C_i = \{(\eta_l)_{1 \leq l \leq N} \in [0, +\infty[^N \mid \forall k \sum_{l \in N_k} \eta_l = d_{ik}\}$
- For every $j \in \{1, \dots, M\}$,
 - $\phi_j: \mathbb{R} \rightarrow [0, +\infty[$: strictly increasing τ -Lipschitz continuous function (cost of transiting on link j).
 - $h_j: \mathbb{R}^{Nm} \rightarrow \mathbb{R}: (x_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m (Lx_i)^\top e_j$ total flow through link j .

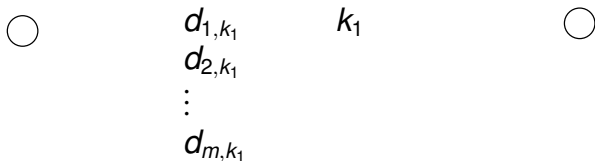
Network Flows

- Directed graph (transport network).
- m types of flows, M links, N paths, Q origin-destination pairs. L is a $M \times N$ binary matrix (L_{jl} is 1 if link j belongs to path l and 0 otherwise).
- For every $i \in \{1, \dots, m\}$, $k \in \{1, \dots, Q\}$
 - $x_i \in \mathbb{R}^N$: flow type i on all the routes.
 - d_{ik} : total flow type i that transit from the origin to the destination of k .
 - N_k : number of paths linking k .
 - $C_i = \{(\eta_l)_{1 \leq l \leq N} \in [0, +\infty[^N \mid \forall k \sum_{l \in N_k} \eta_l = d_{ik}\}$
- For every $j \in \{1, \dots, M\}$,
 - $\phi_j: \mathbb{R} \rightarrow [0, +\infty[$: strictly increasing τ -Lipschitz continuous function (cost of transiting on link j).
 - $h_j: \mathbb{R}^{Nm} \rightarrow \mathbb{R}: (x_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m (Lx_i)^\top e_j$ total flow through link j .

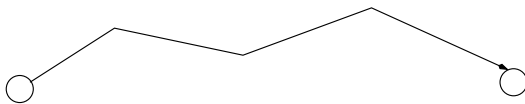
Network Flows

 k_1 

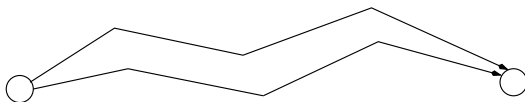
Network Flows



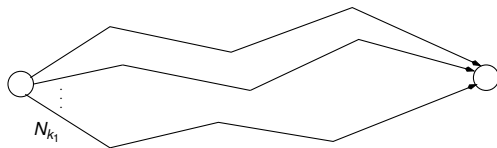
Network Flows



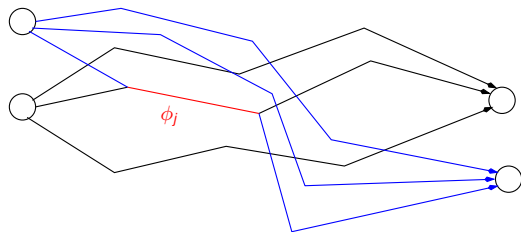
Network Flows



Network Flows



Network Flows



Network Flows

The Wardrop equilibria are the solutions of

Beckmann et al.'56

$$(N) \quad \underset{x_1 \in C_1, \dots, x_m \in C_m}{\text{minimize}} \quad \sum_{j=1}^M \int_0^{h_j(x_1, \dots, x_m)} \phi_j(h) dh.$$

Setting, for every $i \in \{1, \dots, m\}$,

- $A_i = N_{C_i}$
- $B_j: (x_i)_{1 \leq i \leq m} \mapsto L^\top(\phi_1(\sum_{i=1}^m (Lx_i)^\top e_1), \dots, \phi_m(\sum_{i=1}^m (Lx_i)^\top e_m))$

(N) is a particular case of (P). Then our Algorithm, which becomes

$\forall i \quad x_{i,n+1} = P_{C_i}(x_{i,n} - \gamma L^\top(\phi_1(\rho_{1,n}), \dots, \phi_m(\rho_{m,n})))$,
 where $(\rho_{1,n}, \dots, \rho_{m,n}) = \sum_{j=1}^m Lx_{j,n}$, converges (strongly) to the solution of (N).

Network Flows

The Wardrop equilibria are the solutions of

Beckmann et al.'56

$$(N) \quad \underset{x_1 \in C_1, \dots, x_m \in C_m}{\text{minimize}} \quad \sum_{j=1}^M \int_0^{h_j(x_1, \dots, x_m)} \phi_j(h) dh.$$

Setting, for every $i \in \{1, \dots, m\}$,

- $A_i = N_{C_i}$

- $B_i: (x_j)_{1 \leq j \leq m} \mapsto L^\top(\phi_1(\sum_{i=1}^m (Lx_i)^\top e_1), \dots, \phi_m(\sum_{i=1}^m (Lx_i)^\top e_m))$

(N) is a particular case of (P). Then our Algorithm, which becomes

$$\forall i \quad x_{i,n+1} = P_{C_i}(x_{i,n} - \gamma L^\top(\phi_1(\rho_{1,n}), \dots, \phi_m(\rho_{m,n}))),$$

where $(\rho_{1,n}, \dots, \rho_{m,n}) = \sum_{j=1}^m Lx_{j,n}$, converges (strongly) to the solution of (N).

Network Flows

The Wardrop equilibria are the solutions of

Beckmann et al.'56

$$(N) \quad \underset{x_1 \in C_1, \dots, x_m \in C_m}{\text{minimize}} \quad \sum_{j=1}^M \int_0^{h_j(x_1, \dots, x_m)} \phi_j(h) dh.$$

Setting, for every $i \in \{1, \dots, m\}$,

- $A_i = N_{C_i}$

- $B_i: (x_j)_{1 \leq j \leq m} \mapsto L^\top(\phi_1(\sum_{i=1}^m (Lx_i)^\top e_1), \dots, \phi_m(\sum_{i=1}^m (Lx_i)^\top e_m))$

(N) is a particular case of (P). Then our Algorithm, which becomes

$$\forall i \quad x_{i,n+1} = P_{C_i}(x_{i,n} - \gamma L^\top(\phi_1(\rho_{1,n}), \dots, \phi_m(\rho_{m,n}))),$$

where $(\rho_{1,n}, \dots, \rho_{m,n}) = \sum_{j=1}^m Lx_{j,n}$, converges (strongly) to the solution of (N).

Thank you for your attention !