

A Splitting Algorithm for Coupled System of Primal–Dual Monotone Inclusions

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Outline

- 1 Problem statement
- 2 Algorithm
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- 5 Open problems and future work

Extended set of solution

- $\mathbf{A}: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ is maximally monotone on a real Hilbert space \mathcal{H} .
- $\mathbf{C}: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ is maximally monotone.
- Suppose that $0 \in \mathbf{A}\bar{\mathbf{x}} + \mathbf{C}\bar{\mathbf{x}}$ then there exists $\bar{\mathbf{y}} \in \mathbf{C}\bar{\mathbf{x}}$ such that $-\bar{\mathbf{y}} \in \mathbf{A}\bar{\mathbf{x}}$. Define extended set of solution

$$\text{Ext} = \{(\mathbf{x}, \mathbf{y}) \in \mathcal{H}^2 \mid -\mathbf{y} \in \mathbf{A}\mathbf{x} \text{ and } \mathbf{y} \in \mathbf{C}\mathbf{x}\} \neq \emptyset$$

- Examples: Forward-backward, Forward-backward-forward, Douglas-Racford.

Extended set of solution (Breceno-Combettes 2011)

- $\mathbf{B}: \mathcal{G} \rightarrow 2^{\mathcal{G}}$ is maximally monotone on Hilbert space \mathcal{G} .
- $\mathbf{L}: \mathcal{H} \rightarrow \mathcal{G}$ is a linear and bounded.
- Suppose that $0 \in \mathbf{A}\bar{\mathbf{x}} + \mathbf{L}^* \circ \mathbf{B} \circ \mathbf{L}\bar{\mathbf{x}}$. Define extended set of solution

$$\text{Ext} = \{(\mathbf{x}, \mathbf{v}) \in \mathcal{H} \times \mathcal{G} \mid -\mathbf{L}^* \mathbf{v} \in \mathbf{A}\mathbf{x} \text{ and } \mathbf{L}\mathbf{x} \in \mathbf{B}^{-1} \mathbf{v}\} \neq \emptyset$$

- Examples: Primal-Dual Forward-Backward-Forward.

Extended set of solution (Combettes-Pesquet 2011)

- $D: \mathcal{G} \rightarrow 2^{\mathcal{G}}$ is maximally monotone.
- $B \square D = (B^{-1} + D^{-1})^{-1}$.
- Suppose that $0 \in A\bar{x} + L^* \circ (B \square D) \circ L\bar{x} + C\bar{x}$. Define extended set of solution

$$E = \{(\mathbf{x}, \mathbf{v}) \in \mathcal{H} \times \mathcal{G} \mid -L^* \mathbf{v} \in (A + C)\bar{x} \text{ and } L\mathbf{x} \in (B^{-1} + D^{-1})\mathbf{v}\} \neq \emptyset$$

- Examples: Primal-Dual Forward-Backward-Forward.

System of monotone inclusions

- $\mathcal{H} = \mathcal{H}_1 \times \dots \times \mathcal{H}_m$.

- Primal coupling: $C_i: \mathcal{H} \rightarrow \mathcal{H}_i$, $(\mathbf{C}: \mathcal{H} \rightarrow \mathcal{H}: \mathbf{x} \mapsto (\mathbf{C}_i \mathbf{x})_{1 \leq i \leq m})$

$$\langle \mathbf{x} - \mathbf{y} \mid \mathbf{C}\mathbf{x} - \mathbf{C}\mathbf{y} \rangle \geq \nu_0 \|\mathbf{C}\mathbf{x} - \mathbf{C}\mathbf{y}\|^2.$$

- $\mathcal{G} = \mathcal{G}_1 \times \dots \times \mathcal{G}_s$.

- Dual coupling: $S_i: \mathcal{H} \rightarrow \mathcal{H}_i$, $(\mathbf{D}: \mathcal{G} \rightarrow \mathcal{G}: \mathbf{v} \mapsto (\mathbf{S}_i \mathbf{v})_{1 \leq i \leq s})$

$$\langle \mathbf{v} - \mathbf{w} \mid \mathbf{D}\mathbf{v} - \mathbf{D}\mathbf{w} \rangle \geq \mu_0 \|\mathbf{D}\mathbf{v} - \mathbf{D}\mathbf{w}\|^2.$$

System of monotone inclusions

- Linear operators $\mathbf{L} = ((L_{k,i})_{1 \leq k \leq s})_{1 \leq i \leq m}$, where $L_{k,i}: \mathcal{H}_i \rightarrow \mathcal{G}_k$.
- Maximally monotone $\mathbf{A} = \times_{i=1}^m A_i$, where $A_i: \mathcal{H}_i \rightarrow 2^{\mathcal{H}_i}$.
- Maximally monotone $\mathbf{B} = \times_{i=1}^m B_i$, where $B_i: \mathcal{G}_i \rightarrow 2^{\mathcal{G}_i}$.
- Problem:

$$\left\{ \begin{array}{l} \mathbf{z}_1 - \sum_{k=1}^s L_{k,1}^* \bar{\mathbf{v}}_k \in A_1 \bar{\mathbf{x}}_1 + C_1 \bar{\mathbf{x}} \\ \vdots \\ \mathbf{z}_m - \sum_{k=1}^s L_{k,m}^* \bar{\mathbf{v}}_k \in A_m \bar{\mathbf{x}}_m + C_m \bar{\mathbf{x}} \end{array} \right. \quad \text{and} \quad \left\{ \begin{array}{l} \sum_{i=1}^m L_{1,i} \bar{\mathbf{x}}_i - r_1 \in B_1 \bar{\mathbf{v}}_1 + S_1 \bar{\mathbf{v}} \\ \vdots \\ \sum_{i=1}^m L_{s,i} \bar{\mathbf{x}}_i - r_s \in B_s \bar{\mathbf{v}}_s + S_s \bar{\mathbf{v}} \end{array} \right.$$

- Equivalently: $\mathbf{z} - \mathbf{L}^* \bar{\mathbf{v}} \in (\mathbf{A} + \mathbf{C}) \bar{\mathbf{x}}$ and $\mathbf{L} \bar{\mathbf{x}} - \mathbf{r} \in (\mathbf{B} + \mathbf{D}) \bar{\mathbf{v}}$.

Ex1: Attouch- Briceño-Arias-Combettes 2010

- Set $L = ((L_{k,i})_{1 \leq k \leq s})_{1 \leq i \leq m} = 0$.

$$\begin{cases} z_1 \in A_1 \bar{x}_1 + C_1 \bar{x} \\ \vdots \\ z_m \in A_m \bar{x}_m + C_m \bar{x} \end{cases} \quad \text{and} \quad \begin{cases} -r_1 \in B_1 \bar{v}_1 + S_1 \bar{v} \\ \vdots \\ -r_s \in B_s \bar{v}_s + S_s \bar{v}. \end{cases}$$

Ex2: Combettes 2013 $C_i(\mathbf{x}) = C_i(x_i)$ and $S_k(\mathbf{v}) = S_k(v_k)$.

- Primal inclusion ($R_k = S_k^{-1}$), $(A \square B = (A^{-1} + B^{-1})^{-1})$

$$\begin{cases} z_1 \in A_1 \bar{x}_1 + \sum_{k=1}^s L_{k,1}^* \left((R_k \square B_k) \left(\sum_{i=1}^m L_{k,i} \bar{x}_i - r_k \right) \right) + C_1 \bar{x}_1 \\ \vdots \\ z_m \in A_m \bar{x}_m + \sum_{k=1}^s L_{k,m}^* \left((R_k \square B_k) \left(\sum_{i=1}^m L_{k,i} \bar{x}_i - r_k \right) \right) + C_m \bar{x}_m. \end{cases}$$

- Dual inclusion:

$$\begin{cases} -r_1 \in B_1 \bar{v}_1 + \sum_{i=1}^m L_{1,i} \left((A_i \square C_i)^{-1} \left(z_i - \sum_{i=1}^m L_{k,i}^* \bar{v}_i \right) \right) + S_1 \bar{v}_1 \\ \vdots \\ -r_s \in B_s \bar{v}_s + \sum_{i=1}^m L_{s,i} \left((A_i \square C_i)^{-1} \left(z_i - \sum_{i=1}^m L_{k,i}^* \bar{v}_i \right) \right) + S_s \bar{v}_s. \end{cases}$$

Ex3: Boj-Csetnek-Nagy 2013

■ Solve system

$$\begin{cases} 0 \in L_{1,1}^*((R_1 \square B_1)(L_{1,1}\bar{x}_1)) + C_1\bar{x} \\ \vdots \\ 0 \in L_{m,m}^*((R_m \square B_m)(L_{m,m}\bar{x}_m)) + C_m\bar{x}. \end{cases}$$

Some features of the algorithm proposed

- Use the approximations of the resolvents of the set-valued operators: $J_{UA} = (\text{Id} + UA)^{-1}$ (see Combettes-Vu).
- Use the approximations of the single-valued operators C_i and S_k in the sense of Attouch- Briceño-Arias-Combettes 2010.

- $(C_{i,n})_{n \in \mathbb{N}}$ are operators from \mathcal{H} to \mathcal{H}_i such that
 (b1) $C_{i,n} - C_i$ is $\kappa_{i,n}$ -Lipschitz continuous satisfying

$$\sum_{n \in \mathbb{N}} \kappa_{i,n} < +\infty \text{ and } (\exists \bar{\mathbf{s}} \in \mathcal{H})(\forall n \in \mathbb{N}) \quad C_{i,n} \bar{\mathbf{s}} = C_i \bar{\mathbf{s}}.$$

- $(S_{k,n})_{n \in \mathbb{N}}$ are operators from \mathcal{G} to \mathcal{G}_k such that
 (c1) $S_{k,n} - S_k$ is $\eta_{k,n}$ -Lipschitz continuous satisfying

$$\sum_{n \in \mathbb{N}} \eta_{k,n} < +\infty \text{ and } (\exists \bar{\mathbf{w}} \in \mathcal{G})(\forall n \in \mathbb{N}) \quad S_{k,n} \bar{\mathbf{w}} = S_k \bar{\mathbf{w}}.$$

- Allow the metric vary over course of iterations (see Combettes-Vu): $(\forall n \in \mathbb{N}) \ U_{i,n+1} \succeq U_{i,n}$ and $V_{k,n+1} \succeq V_{k,n}$, and

$$\mu := \sup_{n \in \mathbb{N}} \{ \|U_{1,n}\|, \dots, \|U_{m,n}\|, \|V_{1,n}\|, \dots, \|V_{s,n}\| \} < +\infty.$$

Algorithms

■ Write $a_n \approx b_n \iff \sum_{n \in \mathbb{N}} \|b_n - a_n\| < +\infty$.

■ Algorithm

(i) For $i = 1, \dots, m$

$$1. t_{i,n} \approx \sum_{k=1}^s L_{k,i}^* v_{k,n} + C_{i,n}(x_{1,n}, \dots, x_{m,n})$$

$$2. p_{i,n} \approx J_{U_{i,n}A_i}(x_{i,n} - U_{i,n}(t_{i,n} - z_i))$$

$$3. y_{i,n} \approx 2p_{i,n} - x_{i,n}$$

$$4. x_{i,n+1} \approx x_{i,n} + \lambda_n(p_{i,n} - x_{i,n})$$

(ii) For $k = 1, \dots, s$

$$1. w_{k,n} \approx \sum_{i=1}^m L_{k,i} y_{i,n} - S_{k,n}(v_{1,n}, \dots, v_{s,n})$$

$$2. q_{k,n} \approx J_{V_{k,n}B_k}(v_{k,n} + V_{k,n}(w_{k,n} - r_k))$$

$$3. v_{k,n+1} \approx v_{k,n} + \lambda_n(q_{k,n} - v_{k,n}),$$

■ This is a non-trivial extension of the algorithm in Combettes-Vu (2013) from univariate to multi-variate C_i, S_k , from C_i, S_k to $C_{i,n}, S_{k,n}$.

Weak convergence

- Assume that the set of solution is non empty and $\exists L_{k_0, i_0} \neq 0$.
- Define

$$\delta_n = \left(\sqrt{\sum_{i=1}^m \sum_{k=1}^s \left\| \sqrt{V_{k,n}} L_{k,i} \sqrt{U_{i,n}} \right\|^2} \right)^{-1} - 1,$$

- Set $\beta = \min\{\mu_0, \nu_0\}$ and suppose that

$$\zeta_n := \frac{\delta_n}{(1 + \delta_n) \max_{1 \leq i \leq m, 1 \leq k \leq s} \{\|U_{i,n}\|, \|V_{k,n}\|\}} \geq \frac{1}{2\beta - \varepsilon}.$$

- Then, $(x_{1,n}, \dots, x_{m,n}, v_{1,n}, \dots, v_{s,n})_{n \in \mathbb{N}}$ converges strongly to a solution $(\bar{x}_1, \dots, \bar{x}_m, \bar{v}_1, \dots, \bar{v}_s)$.

Strong convergence

- Recall (Attouch-Briceno-Combettes): A set-valued operator A is said to be demiregular at $x \in \text{dom } A$ if for every $v \in Ax$ and

$$\begin{cases} x_n \rightarrow x \\ Ax_n \ni v_n \rightarrow v \end{cases} \implies x_n \rightarrow x.$$

- If \mathbf{C} is demiregular at $(\bar{x}_1, \dots, \bar{x}_m)$, $(x_{1,n}, \dots, x_{m,n}) \rightarrow (\bar{x}_1, \dots, \bar{x}_m)$.
- If \mathbf{D} is demiregular at $(\bar{v}_1, \dots, \bar{v}_s)$, $(v_{1,n}, \dots, v_{s,n}) \rightarrow (\bar{v}_1, \dots, \bar{v}_s)$.
- Suppose that there exists $j \in \{1, \dots, m\}$ and $\mathbf{C}: \mathcal{H}_j \rightarrow \mathcal{H}_j$ such that $(\forall (x_i)_{1 \leq i \leq m} \in (\mathcal{H}_i)_{1 \leq i \leq m}) \mathbf{C}_j(x_1, \dots, x_m) = \mathbf{C}x_j$ and \mathbf{C} is demiregular at \bar{x}_j , then $x_{j,n} \rightarrow \bar{x}_j$.
- Suppose that there exists $j \in \{1, \dots, s\}$ and $\mathbf{D}: \mathcal{G}_j \rightarrow \mathcal{G}_j$ such that $(\forall (v_k)_{1 \leq k \leq s} \in (\mathcal{G}_k)_{1 \leq k \leq s}) \mathbf{S}_j(v_1, \dots, v_s) = \mathbf{D}v_j$ and \mathbf{D} is demiregular at \bar{v}_j , then $v_{j,n} \rightarrow \bar{v}_j$.

non-smooth coupling + smooth coupling + prior information

- Smooth coupling (see also Attouch-Brecheno-Combettes):
 $\varphi: \mathcal{H}_1 \times \dots \times \mathcal{H}_m \rightarrow \mathbb{R}$ is a convex differentiable function with ν_0^{-1} -Lipschitz continuous gradient.
- Non-smooth coupling (Combettes 2013): $\ell_k, \mathbf{g}_k \in \Gamma_0(\mathcal{G}_k)$,
 $\ell_k \square \mathbf{g}_k: \mathbf{x} \mapsto \inf_{\mathbf{y} \in \mathcal{G}} (\ell_k(\mathbf{x} - \mathbf{y}) + \mathbf{g}_k(\mathbf{y}))$

$$\psi: (\mathbf{x}_1, \dots, \mathbf{x}_m) \mapsto \sum_{k=1}^s (\ell_k \square \mathbf{g}_k) \left(\sum_{i=1}^m L_{k,i} \mathbf{x}_i - r_k \right)$$

- f_i model the prior information on $\bar{\mathbf{x}}_i$.
- Primal problem:

$$\underset{\mathbf{x}_1 \in \mathcal{H}_1, \dots, \mathbf{x}_m \in \mathcal{H}_m}{\text{minimize}} \sum_{i=1}^m (f_i(\mathbf{x}_i) - \langle \mathbf{x}_i \mid \mathbf{z}_i \rangle) + \psi(\mathbf{x}_1, \dots, \mathbf{x}_m) + \varphi(\mathbf{x}_1, \dots, \mathbf{x}_m),$$

Non-smooth coupling + smooth coupling + prior information

- $f^* : a \mapsto \sup_x (\langle a | x \rangle - f(x))$.
- Dual problem:

$$\begin{aligned} \text{minimize}_{v_1 \in \mathcal{G}_1, \dots, v_s \in \mathcal{G}_s} & \left(\varphi^* \square \left(\sum_{i=1}^m f_i^* \right) \right) \left(\left(z_i - \sum_{k=1}^s L_{k,i}^* v_k \right)_{1 \leq i \leq m} \right) \\ & + \sum_{k=1}^s \left(\ell_k^*(v_k) + g_k^*(v_k) + \langle v_k | r_k \rangle \right). \end{aligned}$$

- Assumption: $\exists \bar{\mathbf{x}} = (\bar{x}_1, \dots, \bar{x}_m)$ such that, for every $i \in \{1, \dots, m\}$,

$$z_i \in \partial f_i(\bar{x}_i) + \sum_{k=1}^s L_{k,i}^* \circ \left(\partial \ell_k \square \partial g_k \right) \circ \left(\sum_{j=1}^m L_{k,j} \bar{x}_j - r_k \right) + \nabla_i \varphi(\bar{\mathbf{x}}),$$

where $\nabla_i \varphi$ is the *ith* component of the gradient $\nabla \varphi$,

Non-smooth coupling + smooth coupling + prior information

- Operator proximal:

$$J_{U^{-1}\partial f} := \text{prox}_f^U: \mathcal{H} \rightarrow \mathcal{H}: x \mapsto \arg \min_{y \in \mathcal{H}} (f(y) + \frac{1}{2} \|x - y\|_U^2)$$

- $(\varphi_n)_{n \in \mathbb{N}}$ are differentiable functions in $\Gamma_0(\mathcal{H}_1 \times \dots \times \mathcal{H}_m)$ such that

- $\nabla_i \varphi_n - \nabla_i \varphi$ is κ_n Lipschitz continuous satisfying $\sum_{n \in \mathbb{N}} \kappa_{i,n} < +\infty$ and $(\exists \bar{\mathbf{s}} \in \mathcal{H})(\forall n \in \mathbb{N}) \nabla_i \varphi_n(\bar{\mathbf{s}}) = \nabla_i \varphi(\bar{\mathbf{s}})$.

- $(\tilde{\ell}_{k,n})_{n \in \mathbb{N}}$ are differentiable functions in $\Gamma_0(\mathcal{G}_k)$ such that

- $(\nabla \tilde{\ell}_{k,n} - \nabla \ell_k^*)_{n \in \mathbb{N}}$ is $\eta_{k,n}$ Lipschitz continuous satisfying $\sum_{n \in \mathbb{N}} \eta_{k,n} < +\infty$ and $(\exists \bar{\mathbf{w}} := (\bar{w}_j)_{1 \leq j \leq s} \in \mathcal{G})(\forall n \in \mathbb{N}) \nabla \tilde{\ell}_{k,n}(\bar{\mathbf{w}}_k) = \nabla \ell_k^*(\bar{\mathbf{w}}_k)$.

Algorithms

■ Algorithm

(i) For $i = 1, \dots, m$

$$1. \ t_{i,n} \approx \sum_{k=1}^s L_{k,i}^* v_{k,n} + \nabla_i \varphi_n(x_{1,n}, \dots, x_{m,n})$$

$$2. \ p_{i,n} \approx \text{prox}_{f_i}^{U_{i,n}^{-1}}(x_{i,n} - U_{i,n}(t_{i,n} - z_i))$$

$$3. \ y_{i,n} := 2p_{i,n} - x_{i,n}$$

$$4. \ x_{i,n+1} := x_{i,n} + \lambda_n(p_{i,n} - x_{i,n})$$

(ii) For $k = 1, \dots, s$

$$1. \ w_{k,n} \approx \sum_{i=1}^m L_{k,i} y_{i,n} - \nabla \tilde{\ell}_{k,n}(v_{k,n})$$

$$2. \ q_{k,n} \approx \text{prox}_{g_k^*}^{V_{k,n}^{-1}}(v_{k,n} + V_{k,n}(w_{k,n} - r_k))$$

$$3. \ v_{k,n+1} := v_{k,n} + \lambda_n(q_{k,n} - v_{k,n}),$$

Weak convergence

- Assume that the set of solution is non empty and $\exists L_{k_0, i_0} \neq 0$.
- Define

$$\delta_n = \left(\sqrt{\sum_{i=1}^m \sum_{k=1}^s \left\| \sqrt{V_{k,n}} L_{k,i} \sqrt{U_{i,n}} \right\|^2} \right)^{-1} - 1,$$

- Set $\beta = \min\{\mu_0, \nu_0\}$ and suppose that

$$\zeta_n := \frac{\delta_n}{(1 + \delta_n) \max_{1 \leq i \leq m, 1 \leq k \leq s} \{\|U_{i,n}\|, \|V_{k,n}\|\}} \geq \frac{1}{2\beta - \varepsilon}.$$

- Then, $(x_{1,n}, \dots, x_{m,n}, v_{1,n}, \dots, v_{s,n})_{n \in \mathbb{N}}$ converges strongly to a solution $(\bar{x}_1, \dots, \bar{x}_m, \bar{v}_1, \dots, \bar{v}_s)$.

Strong convergence

- If φ is uniformly convex at $(\bar{x}_1, \dots, \bar{x}_m)$, then $(x_{1,n}, \dots, x_{m,n}) \rightarrow (\bar{x}_1, \dots, \bar{x}_m)$.
- If $\varphi(x_1, \dots, x_m) = h_1(x_1) + \dots + h_m(x_m)$ for some convex differentiable h_i and h_j is uniformly convex at \bar{x}_j , then $x_{j,n} \rightarrow \bar{x}_j$.
- Suppose that ℓ_j^* is uniformly convex at \bar{v}_j , for some $j \in \{1, \dots, s\}$, then $v_{j,n} \rightarrow \bar{v}_j$.

Open problem and future work

- Implementation of Algorithm recovery of signal over dictionary with non trivial $U_{i,n}$ and $V_{k,n}$?
- Rate of convergence in expectation of the iterations and of the function-valued for strongly smooth coupling and their unbiased estimations are known (See also, Lorenzo-Silva-Vu, Combettes-Pesquet, Pesquet-Repetti)?
- $U_{i,n}$ and $V_{k,n}$ are random vectors with values in $\mathcal{P}_\alpha(\mathcal{H}_i)$ and $\mathcal{P}_\alpha(\mathcal{G}_k)$, respectively, and with unbiased of smooth coupling. (see Bannar 2006 for the case of stochastic projected gradient)

Main references

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